

wrangle_act

June 29, 2018

```
In [3]: import pandas as pd
import numpy as np
import requests
import tweepy
import json
import re
import matplotlib.pyplot as plt
import warnings
```

1 Gather

```
In [4]: df_twitter = pd.read_csv('twitter-archive-enhanced.csv', encoding='utf-8')
df_twitter
```

```
Out[4]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	
5	891087950875897856	NaN	NaN	
6	890971913173991426	NaN	NaN	
7	890729181411237888	NaN	NaN	
8	890609185150312448	NaN	NaN	
9	890240255349198849	NaN	NaN	
10	890006608113172480	NaN	NaN	
11	889880896479866881	NaN	NaN	
12	889665388333682689	NaN	NaN	
13	889638837579907072	NaN	NaN	
14	889531135344209921	NaN	NaN	
15	889278841981685760	NaN	NaN	
16	888917238123831296	NaN	NaN	
17	888804989199671297	NaN	NaN	
18	888554962724278272	NaN	NaN	
19	888202515573088257	NaN	NaN	
20	888078434458587136	NaN	NaN	
21	887705289381826560	NaN	NaN	

22	887517139158093824	NaN	NaN
23	887473957103951883	NaN	NaN
24	887343217045368832	NaN	NaN
25	887101392804085760	NaN	NaN
26	886983233522544640	NaN	NaN
27	886736880519319552	NaN	NaN
28	886680336477933568	NaN	NaN
29	886366144734445568	NaN	NaN
...
2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2328	666396247373291520	NaN	NaN
2329	666373753744588802	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2332	666345417576210432	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	NaN	NaN
2336	666273097616637952	NaN	NaN
2337	666268910803644416	NaN	NaN
2338	666104133288665088	NaN	NaN
2339	666102155909144576	NaN	NaN
2340	666099513787052032	NaN	NaN
2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2343	666073100786774016	NaN	NaN
2344	666071193221509120	NaN	NaN
2345	666063827256086533	NaN	NaN
2346	666058600524156928	NaN	NaN
2347	666057090499244032	NaN	NaN
2348	666055525042405380	NaN	NaN
2349	666051853826850816	NaN	NaN
2350	666050758794694657	NaN	NaN
2351	666049248165822465	NaN	NaN
2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp \
0	2017-08-01 16:23:56 +0000
1	2017-08-01 00:17:27 +0000
2	2017-07-31 00:18:03 +0000
3	2017-07-30 15:58:51 +0000
4	2017-07-29 16:00:24 +0000
5	2017-07-29 00:08:17 +0000
6	2017-07-28 16:27:12 +0000

7	2017-07-28	00:22:40	+0000
8	2017-07-27	16:25:51	+0000
9	2017-07-26	15:59:51	+0000
10	2017-07-26	00:31:25	+0000
11	2017-07-25	16:11:53	+0000
12	2017-07-25	01:55:32	+0000
13	2017-07-25	00:10:02	+0000
14	2017-07-24	17:02:04	+0000
15	2017-07-24	00:19:32	+0000
16	2017-07-23	00:22:39	+0000
17	2017-07-22	16:56:37	+0000
18	2017-07-22	00:23:06	+0000
19	2017-07-21	01:02:36	+0000
20	2017-07-20	16:49:33	+0000
21	2017-07-19	16:06:48	+0000
22	2017-07-19	03:39:09	+0000
23	2017-07-19	00:47:34	+0000
24	2017-07-18	16:08:03	+0000
25	2017-07-18	00:07:08	+0000
26	2017-07-17	16:17:36	+0000
27	2017-07-16	23:58:41	+0000
28	2017-07-16	20:14:00	+0000
29	2017-07-15	23:25:31	+0000
...			...
2326	2015-11-17	00:24:19	+0000
2327	2015-11-17	00:06:54	+0000
2328	2015-11-16	23:23:41	+0000
2329	2015-11-16	21:54:18	+0000
2330	2015-11-16	21:10:36	+0000
2331	2015-11-16	20:32:58	+0000
2332	2015-11-16	20:01:42	+0000
2333	2015-11-16	19:31:45	+0000
2334	2015-11-16	16:37:02	+0000
2335	2015-11-16	16:11:11	+0000
2336	2015-11-16	15:14:19	+0000
2337	2015-11-16	14:57:41	+0000
2338	2015-11-16	04:02:55	+0000
2339	2015-11-16	03:55:04	+0000
2340	2015-11-16	03:44:34	+0000
2341	2015-11-16	03:22:39	+0000
2342	2015-11-16	02:38:37	+0000
2343	2015-11-16	01:59:36	+0000
2344	2015-11-16	01:52:02	+0000
2345	2015-11-16	01:22:45	+0000
2346	2015-11-16	01:01:59	+0000
2347	2015-11-16	00:55:59	+0000
2348	2015-11-16	00:49:46	+0000
2349	2015-11-16	00:35:11	+0000

2350 2015-11-16 00:30:50 +0000
 2351 2015-11-16 00:24:50 +0000
 2352 2015-11-16 00:04:52 +0000
 2353 2015-11-15 23:21:54 +0000
 2354 2015-11-15 23:05:30 +0000
 2355 2015-11-15 22:32:08 +0000

```

                                source \
0      <a href="http://twitter.com/download/iphone" r...
1      <a href="http://twitter.com/download/iphone" r...
2      <a href="http://twitter.com/download/iphone" r...
3      <a href="http://twitter.com/download/iphone" r...
4      <a href="http://twitter.com/download/iphone" r...
5      <a href="http://twitter.com/download/iphone" r...
6      <a href="http://twitter.com/download/iphone" r...
7      <a href="http://twitter.com/download/iphone" r...
8      <a href="http://twitter.com/download/iphone" r...
9      <a href="http://twitter.com/download/iphone" r...
10     <a href="http://twitter.com/download/iphone" r...
11     <a href="http://twitter.com/download/iphone" r...
12     <a href="http://twitter.com/download/iphone" r...
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14     <a href="http://twitter.com/download/iphone" r...
15     <a href="http://twitter.com/download/iphone" r...
16     <a href="http://twitter.com/download/iphone" r...
17     <a href="http://twitter.com/download/iphone" r...
18     <a href="http://twitter.com/download/iphone" r...
19     <a href="http://twitter.com/download/iphone" r...
20     <a href="http://twitter.com/download/iphone" r...
21     <a href="http://twitter.com/download/iphone" r...
22     <a href="http://twitter.com/download/iphone" r...
23     <a href="http://twitter.com/download/iphone" r...
24     <a href="http://twitter.com/download/iphone" r...
25     <a href="http://twitter.com/download/iphone" r...
26     <a href="http://twitter.com/download/iphone" r...
27     <a href="http://twitter.com/download/iphone" r...
28     <a href="http://twitter.com/download/iphone" r...
29     <a href="http://twitter.com/download/iphone" r...
...
2326  <a href="http://twitter.com/download/iphone" r...
2327  <a href="http://twitter.com/download/iphone" r...
2328  <a href="http://twitter.com/download/iphone" r...
2329  <a href="http://twitter.com/download/iphone" r...
2330  <a href="http://twitter.com/download/iphone" r...
2331  <a href="http://twitter.com/download/iphone" r...
2332  <a href="http://twitter.com/download/iphone" r...
2333  <a href="http://twitter.com/download/iphone" r...
2334  <a href="http://twitter.com/download/iphone" r...

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2335 <a href="http://twitter.com/download/iphone" r...
 2336 <a href="http://twitter.com/download/iphone" r...
 2337 <a href="http://twitter.com/download/iphone" r...
 2338 <a href="http://twitter.com/download/iphone" r...
 2339 <a href="http://twitter.com/download/iphone" r...
 2340 <a href="http://twitter.com/download/iphone" r...
 2341 <a href="http://twitter.com/download/iphone" r...
 2342 <a href="http://twitter.com/download/iphone" r...
 2343 <a href="http://twitter.com/download/iphone" r...
 2344 <a href="http://twitter.com/download/iphone" r...
 2345 <a href="http://twitter.com/download/iphone" r...
 2346 <a href="http://twitter.com/download/iphone" r...
 2347 <a href="http://twitter.com/download/iphone" r...
 2348 <a href="http://twitter.com/download/iphone" r...
 2349 <a href="http://twitter.com/download/iphone" r...
 2350 <a href="http://twitter.com/download/iphone" r...
 2351 <a href="http://twitter.com/download/iphone" r...
 2352 <a href="http://twitter.com/download/iphone" r...
 2353 <a href="http://twitter.com/download/iphone" r...
 2354 <a href="http://twitter.com/download/iphone" r...
 2355 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
0	This is Phineas. He's a mystical boy. Only eve...	NaN
1	This is Tilly. She's just checking pup on you...	NaN
2	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	This is Darla. She commenced a snooze mid meal...	NaN
4	This is Franklin. He would like you to stop ca...	NaN
5	Here we have a majestic great white breaching ...	NaN
6	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	When you watch your owner call another dog a g...	NaN
8	This is Zoey. She doesn't want to be one of th...	NaN
9	This is Cassie. She is a college pup. Studying...	NaN
10	This is Koda. He is a South Australian decksha...	NaN
11	This is Bruno. He is a service shark. Only get...	NaN
12	Here's a puppo that seems to be on the fence a...	NaN
13	This is Ted. He does his best. Sometimes that'...	NaN
14	This is Stuart. He's sporting his favorite fan...	NaN
15	This is Oliver. You're witnessing one of his m...	NaN
16	This is Jim. He found a fren. Taught him how t...	NaN
17	This is Zeke. He has a new stick. Very proud o...	NaN
18	This is Ralphus. He's powering up. Attempting ...	NaN
19	RT @dog_rates: This is Canela. She attempted s...	8.874740e+17
20	This is Gerald. He was just told he didn't get...	NaN
21	This is Jeffrey. He has a monopoly on the pool...	NaN
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
23	This is Canela. She attempted some fancy porch...	NaN
24	You may not have known you needed to see this ...	NaN

25	This... is a Jubilant Antarctic House Bear. We...	NaN
26	This is Maya. She's very shy. Rarely leaves he...	NaN
27	This is Mingus. He's a wonderful father to his...	NaN
28	This is Derek. He's late for a dog meeting. 13...	NaN
29	This is Roscoe. Another pupper fallen victim t...	NaN
...
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2345	This is the happiest dog you will ever see. Ve...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2348	Here is a Siberian heavily armored polar bear ...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2350	This is a truly beautiful English Wilson Staff...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN
2355	Here we have a Japanese Irish Setter. Lost eye...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN
5	NaN	NaN
6	NaN	NaN
7	NaN	NaN
8	NaN	NaN
9	NaN	NaN

10	NaN	NaN
11	NaN	NaN
12	NaN	NaN
13	NaN	NaN
14	NaN	NaN
15	NaN	NaN
16	NaN	NaN
17	NaN	NaN
18	NaN	NaN
19	4.196984e+09	2017-07-19 00:47:34 +0000
20	NaN	NaN
21	NaN	NaN
22	NaN	NaN
23	NaN	NaN
24	NaN	NaN
25	NaN	NaN
26	NaN	NaN
27	NaN	NaN
28	NaN	NaN
29	NaN	NaN
...
2326	NaN	NaN
2327	NaN	NaN
2328	NaN	NaN
2329	NaN	NaN
2330	NaN	NaN
2331	NaN	NaN
2332	NaN	NaN
2333	NaN	NaN
2334	NaN	NaN
2335	NaN	NaN
2336	NaN	NaN
2337	NaN	NaN
2338	NaN	NaN
2339	NaN	NaN
2340	NaN	NaN
2341	NaN	NaN
2342	NaN	NaN
2343	NaN	NaN
2344	NaN	NaN
2345	NaN	NaN
2346	NaN	NaN
2347	NaN	NaN
2348	NaN	NaN
2349	NaN	NaN
2350	NaN	NaN
2351	NaN	NaN
2352	NaN	NaN

2353	NaN	NaN
2354	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator	\
0	https://twitter.com/dog_rates/status/892420643...	13	
1	https://twitter.com/dog_rates/status/892177421...	13	
2	https://twitter.com/dog_rates/status/891815181...	12	
3	https://twitter.com/dog_rates/status/891689557...	13	
4	https://twitter.com/dog_rates/status/891327558...	12	
5	https://twitter.com/dog_rates/status/891087950...	13	
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	13	
7	https://twitter.com/dog_rates/status/890729181...	13	
8	https://twitter.com/dog_rates/status/890609185...	13	
9	https://twitter.com/dog_rates/status/890240255...	14	
10	https://twitter.com/dog_rates/status/890006608...	13	
11	https://twitter.com/dog_rates/status/889880896...	13	
12	https://twitter.com/dog_rates/status/889665388...	13	
13	https://twitter.com/dog_rates/status/889638837...	12	
14	https://twitter.com/dog_rates/status/889531135...	13	
15	https://twitter.com/dog_rates/status/889278841...	13	
16	https://twitter.com/dog_rates/status/888917238...	12	
17	https://twitter.com/dog_rates/status/888804989...	13	
18	https://twitter.com/dog_rates/status/888554962...	13	
19	https://twitter.com/dog_rates/status/887473957...	13	
20	https://twitter.com/dog_rates/status/888078434...	12	
21	https://twitter.com/dog_rates/status/887705289...	13	
22	https://twitter.com/dog_rates/status/887517139...	14	
23	https://twitter.com/dog_rates/status/887473957...	13	
24	https://twitter.com/dog_rates/status/887343217...	13	
25	https://twitter.com/dog_rates/status/887101392...	12	
26	https://twitter.com/dog_rates/status/886983233...	13	
27	https://www.gofundme.com/mingusneedsus,https:/...	13	
28	https://twitter.com/dog_rates/status/886680336...	13	
29	https://twitter.com/dog_rates/status/886366144...	12	
...	
2326	https://twitter.com/dog_rates/status/666411507...	2	
2327	https://twitter.com/dog_rates/status/666407126...	7	
2328	https://twitter.com/dog_rates/status/666396247...	9	
2329	https://twitter.com/dog_rates/status/666373753...	11	
2330	https://twitter.com/dog_rates/status/666362758...	6	
2331	https://twitter.com/dog_rates/status/666353288...	8	
2332	https://twitter.com/dog_rates/status/666345417...	10	
2333	https://twitter.com/dog_rates/status/666337882...	9	
2334	https://twitter.com/dog_rates/status/666293911...	3	
2335	https://twitter.com/dog_rates/status/666287406...	1	
2336	https://twitter.com/dog_rates/status/666273097...	11	
2337	https://twitter.com/dog_rates/status/666268910...	10	

2338	https://twitter.com/dog_rates/status/666104133...	1
2339	https://twitter.com/dog_rates/status/666102155...	11
2340	https://twitter.com/dog_rates/status/666099513...	8
2341	https://twitter.com/dog_rates/status/666094000...	9
2342	https://twitter.com/dog_rates/status/666082916...	6
2343	https://twitter.com/dog_rates/status/666073100...	10
2344	https://twitter.com/dog_rates/status/666071193...	9
2345	https://twitter.com/dog_rates/status/666063827...	10
2346	https://twitter.com/dog_rates/status/666058600...	8
2347	https://twitter.com/dog_rates/status/666057090...	9
2348	https://twitter.com/dog_rates/status/666055525...	10
2349	https://twitter.com/dog_rates/status/666051853...	2
2350	https://twitter.com/dog_rates/status/666050758...	10
2351	https://twitter.com/dog_rates/status/666049248...	5
2352	https://twitter.com/dog_rates/status/666044226...	6
2353	https://twitter.com/dog_rates/status/666033412...	9
2354	https://twitter.com/dog_rates/status/666029285...	7
2355	https://twitter.com/dog_rates/status/666020888...	8

	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None
1	10	Tilly	None	None	None	None
2	10	Archie	None	None	None	None
3	10	Darla	None	None	None	None
4	10	Franklin	None	None	None	None
5	10	None	None	None	None	None
6	10	Jax	None	None	None	None
7	10	None	None	None	None	None
8	10	Zoey	None	None	None	None
9	10	Cassie	doggo	None	None	None
10	10	Koda	None	None	None	None
11	10	Bruno	None	None	None	None
12	10	None	None	None	None	puppo
13	10	Ted	None	None	None	None
14	10	Stuart	None	None	None	puppo
15	10	Oliver	None	None	None	None
16	10	Jim	None	None	None	None
17	10	Zeke	None	None	None	None
18	10	Ralphus	None	None	None	None
19	10	Canela	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None

28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2345	10	the	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2348	10	a	None	None	None	None
2349	10	an	None	None	None	None
2350	10	a	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[2356 rows x 17 columns]

```
In [3]: url = "https://d17h27t6h515a5.cloudfront.net/topher/2017/August/599fd2ad_image-predictions-2017-08-17.jpg"
        response = requests.get(url)
```

```
In [4]: with open('image_predictions.tsv', 'wb') as file:
        file.write(response.content)
```

```
In [5]: df_image = pd.read_csv('image_predictions.tsv', sep='\t')
        df_image
```

```
Out[5]:
```

	tweet_id	jpg_url
0	666020888022790149	https://pbs.twimg.com/media/CT4udnOWwAA0aMy.jpg
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg

3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg
5	666050758794694657	https://pbs.twimg.com/media/CT5Jof1WUAEuVxN.jpg
6	666051853826850816	https://pbs.twimg.com/media/CT5KoJ1WoAAJash.jpg
7	666055525042405380	https://pbs.twimg.com/media/CT5N9tpXIAAifs1.jpg
8	666057090499244032	https://pbs.twimg.com/media/CT5PY90WoAAQGLo.jpg
9	666058600524156928	https://pbs.twimg.com/media/CT5Qw94XAAA_2dP.jpg
10	666063827256086533	https://pbs.twimg.com/media/CT5Vg_wXIAAXfnj.jpg
11	666071193221509120	https://pbs.twimg.com/media/CT5cN_3WEAA10oZ.jpg
12	666073100786774016	https://pbs.twimg.com/media/CT5d9DZXAAALcwe.jpg
13	666082916733198337	https://pbs.twimg.com/media/CT5m4VGWEAAAtKc8.jpg
14	666094000022159362	https://pbs.twimg.com/media/CT5w9gUW4AAsBNN.jpg
15	666099513787052032	https://pbs.twimg.com/media/CT51-JJUEAA6hV8.jpg
16	666102155909144576	https://pbs.twimg.com/media/CT54YGiwUAEZnoK.jpg
17	666104133288665088	https://pbs.twimg.com/media/CT56LSZW0AA1Jj2.jpg
18	666268910803644416	https://pbs.twimg.com/media/CT8QCd1WEAADXws.jpg
19	666273097616637952	https://pbs.twimg.com/media/CT8T1mtUwAA3aqm.jpg
20	666287406224695296	https://pbs.twimg.com/media/CT8g3BpUEAAuFjg.jpg
21	666293911632134144	https://pbs.twimg.com/media/CT8mx7KW4AEQu8N.jpg
22	666337882303524864	https://pbs.twimg.com/media/CT90wFIWEAMuRje.jpg
23	666345417576210432	https://pbs.twimg.com/media/CT9Vn7PW0AA_ZCM.jpg
24	666353288456101888	https://pbs.twimg.com/media/CT9cx0tUEAAhNN_.jpg
25	666362758909284353	https://pbs.twimg.com/media/CT91XGsUcAAyUft.jpg
26	666373753744588802	https://pbs.twimg.com/media/CT9vZEYWUAA1Z05.jpg
27	666396247373291520	https://pbs.twimg.com/media/CT-D2ZHWIAA3gK1.jpg
28	666407126856765440	https://pbs.twimg.com/media/CT-NvwmW4AAAugGZ.jpg
29	666411507551481857	https://pbs.twimg.com/media/CT-RugiWIAELEAq.jpg
...
2045	886366144734445568	https://pbs.twimg.com/media/DE0BTnQUwAApKEH.jpg
2046	886680336477933568	https://pbs.twimg.com/media/DE4fEDzWAAAyHMM.jpg
2047	886736880519319552	https://pbs.twimg.com/media/DE5Se8FXcAAJFx4.jpg
2048	886983233522544640	https://pbs.twimg.com/media/DE8yicJW0AAAvBJ.jpg
2049	887101392804085760	https://pbs.twimg.com/media/DE-eAq6UwAA-jaE.jpg
2050	887343217045368832	https://pbs.twimg.com/ext_tw_video_thumb/88734...
2051	887473957103951883	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2052	887517139158093824	https://pbs.twimg.com/ext_tw_video_thumb/88751...
2053	887705289381826560	https://pbs.twimg.com/media/DFHQBbXgAEqY7t.jpg
2054	888078434458587136	https://pbs.twimg.com/media/DFMWn56WsAAkA7B.jpg
2055	888202515573088257	https://pbs.twimg.com/media/DFDw2tyUQAAAFke.jpg
2056	888554962724278272	https://pbs.twimg.com/media/DFTH_0-UQAACu20.jpg
2057	888804989199671297	https://pbs.twimg.com/media/DFWra-3VYAA2piG.jpg
2058	888917238123831296	https://pbs.twimg.com/media/DFYRgsOUQAARGh0.jpg
2059	889278841981685760	https://pbs.twimg.com/ext_tw_video_thumb/88927...
2060	889531135344209921	https://pbs.twimg.com/media/DFg_2PVW0AEHN3p.jpg
2061	889638837579907072	https://pbs.twimg.com/media/DFihzFfXsAYGDPR.jpg
2062	889665388333682689	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg
2063	889880896479866881	https://pbs.twimg.com/media/DF199B1WsAITKsg.jpg
2064	890006608113172480	https://pbs.twimg.com/media/DFnwSY4WAAAMliS.jpg

2065	890240255349198849	https://pbs.twimg.com/media/DFrEyVuW0AA03t9.jpg
2066	890609185150312448	https://pbs.twimg.com/media/DFwUU_XcAEpyXI.jpg
2067	890729181411237888	https://pbs.twimg.com/media/DFyBahAVwAAhUTd.jpg
2068	890971913173991426	https://pbs.twimg.com/media/DF1eOmZXUAAALUcq.jpg
2069	891087950875897856	https://pbs.twimg.com/media/DF3HwyEWsAABqE6.jpg
2070	891327558926688256	https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.jpg
2071	891689557279858688	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg
2072	891815181378084864	https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg
2073	892177421306343426	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg
2074	892420643555336193	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg

	img_num		p1	p1_conf	p1_dog	\
0	1	Welsh_springer_spaniel	0.465074		True	
1	1	redbone	0.506826		True	
2	1	German_shepherd	0.596461		True	
3	1	Rhodesian_ridgeback	0.408143		True	
4	1	miniature_pinscher	0.560311		True	
5	1	Bernese_mountain_dog	0.651137		True	
6	1	box_turtle	0.933012		False	
7	1	chow	0.692517		True	
8	1	shopping_cart	0.962465		False	
9	1	miniature_poodle	0.201493		True	
10	1	golden_retriever	0.775930		True	
11	1	Gordon_setter	0.503672		True	
12	1	Walker_hound	0.260857		True	
13	1	pug	0.489814		True	
14	1	bloodhound	0.195217		True	
15	1	Lhasa	0.582330		True	
16	1	English_setter	0.298617		True	
17	1	hen	0.965932		False	
18	1	desktop_computer	0.086502		False	
19	1	Italian_greyhound	0.176053		True	
20	1	Maltese_dog	0.857531		True	
21	1	three-toed_sloth	0.914671		False	
22	1	ox	0.416669		False	
23	1	golden_retriever	0.858744		True	
24	1	malamute	0.336874		True	
25	1	guinea_pig	0.996496		False	
26	1	soft-coated_wheaten_terrier	0.326467		True	
27	1	Chihuahua	0.978108		True	
28	1	black-and-tan_coonhound	0.529139		True	
29	1	coho	0.404640		False	
...	
2045	1	French_bulldog	0.999201		True	
2046	1	convertible	0.738995		False	
2047	1	kuvasz	0.309706		True	
2048	2	Chihuahua	0.793469		True	
2049	1	Samoyed	0.733942		True	

2050	1	Mexican_hairless	0.330741	True
2051	2	Pembroke	0.809197	True
2052	1	limousine	0.130432	False
2053	1	basset	0.821664	True
2054	1	French_bulldog	0.995026	True
2055	2	Pembroke	0.809197	True
2056	3	Siberian_husky	0.700377	True
2057	1	golden_retriever	0.469760	True
2058	1	golden_retriever	0.714719	True
2059	1	whippet	0.626152	True
2060	1	golden_retriever	0.953442	True
2061	1	French_bulldog	0.991650	True
2062	1	Pembroke	0.966327	True
2063	1	French_bulldog	0.377417	True
2064	1	Samoyed	0.957979	True
2065	1	Pembroke	0.511319	True
2066	1	Irish_terrier	0.487574	True
2067	2	Pomeranian	0.566142	True
2068	1	Appenzeller	0.341703	True
2069	1	Chesapeake_Bay_retriever	0.425595	True
2070	2	basset	0.555712	True
2071	1	paper_towel	0.170278	False
2072	1	Chihuahua	0.716012	True
2073	1	Chihuahua	0.323581	True
2074	1	orange	0.097049	False

	p2	p2_conf	p2_dog	p3 \
0	collie	0.156665	True	Shetland_sheepdog
1	miniature_pinscher	0.074192	True	Rhodesian_ridgeback
2	malinois	0.138584	True	bloodhound
3	redbone	0.360687	True	miniature_pinscher
4	Rottweiler	0.243682	True	Doberman
5	English_springer	0.263788	True	Greater_Swiss_Mountain_dog
6	mud_turtle	0.045885	False	terrapin
7	Tibetan_mastiff	0.058279	True	fur_coat
8	shopping_basket	0.014594	False	golden_retriever
9	komondor	0.192305	True	soft-coated_wheaten_terrier
10	Tibetan_mastiff	0.093718	True	Labrador_retriever
11	Yorkshire_terrier	0.174201	True	Pekinese
12	English_foxhound	0.175382	True	Ibizan_hound
13	bull_mastiff	0.404722	True	French_bulldog
14	German_shepherd	0.078260	True	malinois
15	Shih-Tzu	0.166192	True	Dandie_Dinmont
16	Newfoundland	0.149842	True	borzoi
17	cock	0.033919	False	partridge
18	desk	0.085547	False	bookcase
19	toy_terrier	0.111884	True	basenji
20	toy_poodle	0.063064	True	miniature_poodle

21	otter	0.015250	False	great_grey_owl
22	Newfoundland	0.278407	True	groenendael
23	Chesapeake_Bay_retriever	0.054787	True	Labrador_retriever
24	Siberian_husky	0.147655	True	Eskimo_dog
25	skunk	0.002402	False	hamster
26	Afghan_hound	0.259551	True	briard
27	toy_terrier	0.009397	True	papillon
28	bloodhound	0.244220	True	flat-coated_retriever
29	barracouta	0.271485	False	gar
...
2045	Chihuahua	0.000361	True	Boston_bull
2046	sports_car	0.139952	False	car_wheel
2047	Great_Pyrenees	0.186136	True	Dandie_Dinmont
2048	toy_terrier	0.143528	True	can_opener
2049	Eskimo_dog	0.035029	True	Staffordshire_bullterrier
2050	sea_lion	0.275645	False	Weimaraner
2051	Rhodesian_ridgeback	0.054950	True	beagle
2052	tow_truck	0.029175	False	shopping_cart
2053	redbone	0.087582	True	Weimaraner
2054	pug	0.000932	True	bull_mastiff
2055	Rhodesian_ridgeback	0.054950	True	beagle
2056	Eskimo_dog	0.166511	True	malamute
2057	Labrador_retriever	0.184172	True	English_setter
2058	Tibetan_mastiff	0.120184	True	Labrador_retriever
2059	borzoi	0.194742	True	Saluki
2060	Labrador_retriever	0.013834	True	redbone
2061	boxer	0.002129	True	Staffordshire_bullterrier
2062	Cardigan	0.027356	True	basenji
2063	Labrador_retriever	0.151317	True	muzzle
2064	Pomeranian	0.013884	True	chow
2065	Cardigan	0.451038	True	Chihuahua
2066	Irish_setter	0.193054	True	Chesapeake_Bay_retriever
2067	Eskimo_dog	0.178406	True	Pembroke
2068	Border_collie	0.199287	True	ice_lolly
2069	Irish_terrier	0.116317	True	Indian_elephant
2070	English_springer	0.225770	True	German_short-haired_pointer
2071	Labrador_retriever	0.168086	True	spatula
2072	malamute	0.078253	True	kelpie
2073	Pekinese	0.090647	True	papillon
2074	bagel	0.085851	False	banana

	p3_conf	p3_dog
0	0.061428	True
1	0.072010	True
2	0.116197	True
3	0.222752	True
4	0.154629	True
5	0.016199	True

6	0.017885	False
7	0.054449	False
8	0.007959	True
9	0.082086	True
10	0.072427	True
11	0.109454	True
12	0.097471	True
13	0.048960	True
14	0.075628	True
15	0.089688	True
16	0.133649	True
17	0.000052	False
18	0.079480	False
19	0.111152	True
20	0.025581	True
21	0.013207	False
22	0.102643	True
23	0.014241	True
24	0.093412	True
25	0.000461	False
26	0.206803	True
27	0.004577	True
28	0.173810	True
29	0.189945	False
...
2045	0.000076	True
2046	0.044173	False
2047	0.086346	True
2048	0.032253	False
2049	0.029705	True
2050	0.134203	True
2051	0.038915	True
2052	0.026321	False
2053	0.026236	True
2054	0.000903	True
2055	0.038915	True
2056	0.111411	True
2057	0.073482	True
2058	0.105506	True
2059	0.027351	True
2060	0.007958	True
2061	0.001498	True
2062	0.004633	True
2063	0.082981	False
2064	0.008167	True
2065	0.029248	True
2066	0.118184	True
2067	0.076507	True

```

2068  0.193548  False
2069  0.076902  False
2070  0.175219   True
2071  0.040836  False
2072  0.031379   True
2073  0.068957   True
2074  0.076110  False

```

```
[2075 rows x 12 columns]
```

```

In [4]: consumer_key = 'Consumer Key'
        consumer_secret = 'Consumer Secret'
        access_token = 'Access Token'
        access_secret = 'Access Secret'

In [5]: auth = tweepy.OAuthHandler(consumer_key, consumer_secret)
        auth.set_access_token(access_token, access_secret)
        api = tweepy.API(auth, parser = tweepy.parsers.JSONParser(), wait_on_rate_limit = True, w

In [6]: # Liste where we will store the dictionaries of our result
        df_list = []
        # Liste frame where we will store the tweet_id of the errors
        error_list = []

        # Get the tweet object for all the tweets in archive dataframe
        for tweet_id in df_twitter['tweet_id']:
            try:
                page = api.get_status(tweet_id, tweet_mode = 'extended')
                # Print one page to look at the structure of the returned file
                # and the names of attributes
                # print(json.dumps(page, indent = 4))
                #break

                favorites = page['favorite_count'] # How many favorites the tweet had
                retweets = page['retweet_count'] # Count of the retweet
                user_followers = page['user']['followers_count'] # How many followers the user h
                user_favourites = page['user']['favourites_count'] # How many favorites the user
                date_time = page['created_at'] # The date and time of the creation

                df_list.append({'tweet_id': int(tweet_id),
                               'favorites': int(favorites),
                               'retweets': int(retweets),
                               'user_followers': int(user_followers),
                               'user_favourites': int(user_favourites),
                               'date_time': pd.to_datetime(date_time)})

            # Catch the exceptions of the TweepError

```



```

except Exception as e:
    print(str(tweet_id)+ " _ " + str(e))
    error_list.append(tweet_id)

888202515573088257 _ [{'code': 144, 'message': 'No status found with that ID.'}]
873697596434513921 _ [{'code': 144, 'message': 'No status found with that ID.'}]
869988702071779329 _ [{'code': 144, 'message': 'No status found with that ID.'}]
866816280283807744 _ [{'code': 144, 'message': 'No status found with that ID.'}]
861769973181624320 _ [{'code': 144, 'message': 'No status found with that ID.'}]
845459076796616705 _ [{'code': 144, 'message': 'No status found with that ID.'}]
842892208864923648 _ [{'code': 144, 'message': 'No status found with that ID.'}]
837012587749474308 _ [{'code': 144, 'message': 'No status found with that ID.'}]
827228250799742977 _ [{'code': 144, 'message': 'No status found with that ID.'}]
802247111496568832 _ [{'code': 144, 'message': 'No status found with that ID.'}]
775096608509886464 _ [{'code': 144, 'message': 'No status found with that ID.'}]
Rate limit reached. Sleeping for: 735
Rate limit reached. Sleeping for: 734

```

```

In [7]: ee_list = []
        for e in error_list:
            try:
                favorites = page['favorite_count']
                retweets = page['retweet_count']
                user_followers = page['user']['followers_count']
                user_favourites = page['user']['favourites_count']
                date_time = page['created_at']

                df_list.append({'tweet_id': int(tweet_id),
                               'favorites': int(favorites),
                               'retweets': int(retweets),
                               'user_followers': int(user_followers),
                               'user_favourites': int(user_favourites),
                               'date_time': pd.to_datetime(date_time)})

            except Exception:
                print(str(tweet_id)+ " _ " + str(e))
                ee_list.append(e)

```

```

In [8]: json_tweets = pd.DataFrame(df_list, columns = ['tweet_id', 'favorites', 'retweets',
                                                       'user_followers', 'user_favourites', 'date_time'])

```

```

# Save the dataframe in file
json_tweets.to_csv('tweet_json.txt', encoding = 'utf-8', index=False)

```

```

In [6]: json_tweets = pd.read_csv('tweet_json.txt', encoding = 'utf-8')
        json_tweets

```

```

Out[6]:
   tweet_id  favorites  retweets  user_followers \
0    892420643555336193    38690     8558      6992402

```

1	892177421306343426	33159	6291	6992402
2	891815181378084864	24962	4174	6992402
3	891689557279858688	42071	8690	6992402
4	891327558926688256	40227	9450	6992402
5	891087950875897856	20170	3128	6992402
6	890971913173991426	11814	2081	6992402
7	890729181411237888	65354	18979	6992402
8	890609185150312448	27724	4280	6992402
9	890240255349198849	31861	7446	6992402
10	890006608113172480	30582	7364	6992402
11	889880896479866881	27722	4990	6992403
12	889665388333682689	48011	10108	6992403
13	889638837579907072	27109	4565	6992403
14	889531135344209921	15058	2243	6992403
15	889278841981685760	25242	5448	6992403
16	888917238123831296	29002	4516	6992403
17	888804989199671297	25537	4362	6992403
18	888554962724278272	19854	3596	6992403
19	888078434458587136	21702	3509	6992403
20	887705289381826560	30109	5415	6992403
21	887517139158093824	46122	11717	6992403
22	887473957103951883	68942	18308	6992403
23	887343217045368832	33596	10448	6992403
24	887101392804085760	30453	5987	6992403
25	886983233522544640	35066	7808	6992403
26	886736880519319552	12039	3312	6992403
27	886680336477933568	22361	4488	6992403
28	886366144734445568	21146	3207	6992403
29	886267009285017600	116	4	6992403
...
2326	666268910803644416	104	35	6992426
2327	666104133288665088	14346	6635	6992426
2328	666102155909144576	80	13	6992426
2329	666099513787052032	156	68	6992426
2330	666094000022159362	164	74	6992426
2331	666082916733198337	119	45	6992426
2332	666073100786774016	322	164	6992426
2333	666071193221509120	148	62	6992426
2334	666063827256086533	476	220	6992426
2335	666058600524156928	112	57	6992426
2336	666057090499244032	298	142	6992426
2337	666055525042405380	434	252	6992426
2338	666051853826850816	1224	853	6992426
2339	666050758794694657	132	58	6992426
2340	666049248165822465	109	41	6992426
2341	666044226329800704	299	141	6992426
2342	666033412701032449	125	45	6992426
2343	666029285002620928	129	47	6992426

2344	666020888022790149	2563	516	6992426
2345	666020888022790149	2563	516	6992426
2346	666020888022790149	2563	516	6992426
2347	666020888022790149	2563	516	6992426
2348	666020888022790149	2563	516	6992426
2349	666020888022790149	2563	516	6992426
2350	666020888022790149	2563	516	6992426
2351	666020888022790149	2563	516	6992426
2352	666020888022790149	2563	516	6992426
2353	666020888022790149	2563	516	6992426
2354	666020888022790149	2563	516	6992426
2355	666020888022790149	2563	516	6992426

	user_favourites	date_time
0	134549	2017-08-01 16:23:56
1	134549	2017-08-01 00:17:27
2	134549	2017-07-31 00:18:03
3	134549	2017-07-30 15:58:51
4	134549	2017-07-29 16:00:24
5	134549	2017-07-29 00:08:17
6	134549	2017-07-28 16:27:12
7	134549	2017-07-28 00:22:40
8	134549	2017-07-27 16:25:51
9	134549	2017-07-26 15:59:51
10	134549	2017-07-26 00:31:25
11	134549	2017-07-25 16:11:53
12	134549	2017-07-25 01:55:32
13	134549	2017-07-25 00:10:02
14	134549	2017-07-24 17:02:04
15	134549	2017-07-24 00:19:32
16	134549	2017-07-23 00:22:39
17	134549	2017-07-22 16:56:37
18	134549	2017-07-22 00:23:06
19	134549	2017-07-20 16:49:33
20	134549	2017-07-19 16:06:48
21	134549	2017-07-19 03:39:09
22	134549	2017-07-19 00:47:34
23	134549	2017-07-18 16:08:03
24	134549	2017-07-18 00:07:08
25	134549	2017-07-17 16:17:36
26	134549	2017-07-16 23:58:41
27	134549	2017-07-16 20:14:00
28	134549	2017-07-15 23:25:31
29	134549	2017-07-15 16:51:35
...
2326	134549	2015-11-16 14:57:41
2327	134549	2015-11-16 04:02:55
2328	134549	2015-11-16 03:55:04

2329	134549	2015-11-16 03:44:34
2330	134549	2015-11-16 03:22:39
2331	134549	2015-11-16 02:38:37
2332	134549	2015-11-16 01:59:36
2333	134549	2015-11-16 01:52:02
2334	134549	2015-11-16 01:22:45
2335	134549	2015-11-16 01:01:59
2336	134549	2015-11-16 00:55:59
2337	134549	2015-11-16 00:49:46
2338	134549	2015-11-16 00:35:11
2339	134549	2015-11-16 00:30:50
2340	134549	2015-11-16 00:24:50
2341	134549	2015-11-16 00:04:52
2342	134549	2015-11-15 23:21:54
2343	134549	2015-11-15 23:05:30
2344	134549	2015-11-15 22:32:08
2345	134549	2015-11-15 22:32:08
2346	134549	2015-11-15 22:32:08
2347	134549	2015-11-15 22:32:08
2348	134549	2015-11-15 22:32:08
2349	134549	2015-11-15 22:32:08
2350	134549	2015-11-15 22:32:08
2351	134549	2015-11-15 22:32:08
2352	134549	2015-11-15 22:32:08
2353	134549	2015-11-15 22:32:08
2354	134549	2015-11-15 22:32:08
2355	134549	2015-11-15 22:32:08

[2356 rows x 6 columns]

2 Access

In [12]: df_twitter.head()

Out[12]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
0	892420643555336193	NaN	NaN	
1	892177421306343426	NaN	NaN	
2	891815181378084864	NaN	NaN	
3	891689557279858688	NaN	NaN	
4	891327558926688256	NaN	NaN	

	timestamp	\
0	2017-08-01 16:23:56 +0000	
1	2017-08-01 00:17:27 +0000	
2	2017-07-31 00:18:03 +0000	
3	2017-07-30 15:58:51 +0000	
4	2017-07-29 16:00:24 +0000	

	source \		text	retweeted_status_id \
0	<a href="http://twitter.com/download/iphone" r...		This is Phineas. He's a mystical boy. Only eve...	NaN
1	<a href="http://twitter.com/download/iphone" r...		This is Tilly. She's just checking pup on you...	NaN
2	<a href="http://twitter.com/download/iphone" r...		This is Archie. He is a rare Norwegian Pouncin...	NaN
3	<a href="http://twitter.com/download/iphone" r...		This is Darla. She commenced a snooze mid meal...	NaN
4	<a href="http://twitter.com/download/iphone" r...		This is Franklin. He would like you to stop ca...	NaN

	retweeted_status_user_id	retweeted_status_timestamp \
0	NaN	NaN
1	NaN	NaN
2	NaN	NaN
3	NaN	NaN
4	NaN	NaN

	expanded_urls	rating_numerator \
0	https://twitter.com/dog_rates/status/892420643...	13
1	https://twitter.com/dog_rates/status/892177421...	13
2	https://twitter.com/dog_rates/status/891815181...	12
3	https://twitter.com/dog_rates/status/891689557...	13
4	https://twitter.com/dog_rates/status/891327558...	12

	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None
1	10	Tilly	None	None	None	None
2	10	Archie	None	None	None	None
3	10	Darla	None	None	None	None
4	10	Franklin	None	None	None	None

In [7]: df_twitter

Out[7]:

	tweet_id	in_reply_to_status_id	in_reply_to_user_id \
0	892420643555336193	NaN	NaN
1	892177421306343426	NaN	NaN
2	891815181378084864	NaN	NaN
3	891689557279858688	NaN	NaN
4	891327558926688256	NaN	NaN
5	891087950875897856	NaN	NaN
6	890971913173991426	NaN	NaN
7	890729181411237888	NaN	NaN
8	890609185150312448	NaN	NaN

9	890240255349198849	NaN	NaN
10	890006608113172480	NaN	NaN
11	889880896479866881	NaN	NaN
12	889665388333682689	NaN	NaN
13	889638837579907072	NaN	NaN
14	889531135344209921	NaN	NaN
15	889278841981685760	NaN	NaN
16	888917238123831296	NaN	NaN
17	888804989199671297	NaN	NaN
18	888554962724278272	NaN	NaN
19	888202515573088257	NaN	NaN
20	888078434458587136	NaN	NaN
21	887705289381826560	NaN	NaN
22	887517139158093824	NaN	NaN
23	887473957103951883	NaN	NaN
24	887343217045368832	NaN	NaN
25	887101392804085760	NaN	NaN
26	886983233522544640	NaN	NaN
27	886736880519319552	NaN	NaN
28	886680336477933568	NaN	NaN
29	886366144734445568	NaN	NaN
...
2326	666411507551481857	NaN	NaN
2327	666407126856765440	NaN	NaN
2328	666396247373291520	NaN	NaN
2329	666373753744588802	NaN	NaN
2330	666362758909284353	NaN	NaN
2331	666353288456101888	NaN	NaN
2332	666345417576210432	NaN	NaN
2333	666337882303524864	NaN	NaN
2334	666293911632134144	NaN	NaN
2335	666287406224695296	NaN	NaN
2336	666273097616637952	NaN	NaN
2337	666268910803644416	NaN	NaN
2338	666104133288665088	NaN	NaN
2339	666102155909144576	NaN	NaN
2340	666099513787052032	NaN	NaN
2341	666094000022159362	NaN	NaN
2342	666082916733198337	NaN	NaN
2343	666073100786774016	NaN	NaN
2344	666071193221509120	NaN	NaN
2345	666063827256086533	NaN	NaN
2346	666058600524156928	NaN	NaN
2347	666057090499244032	NaN	NaN
2348	666055525042405380	NaN	NaN
2349	666051853826850816	NaN	NaN
2350	666050758794694657	NaN	NaN
2351	666049248165822465	NaN	NaN

2352	666044226329800704	NaN	NaN
2353	666033412701032449	NaN	NaN
2354	666029285002620928	NaN	NaN
2355	666020888022790149	NaN	NaN

	timestamp \
0	2017-08-01 16:23:56 +0000
1	2017-08-01 00:17:27 +0000
2	2017-07-31 00:18:03 +0000
3	2017-07-30 15:58:51 +0000
4	2017-07-29 16:00:24 +0000
5	2017-07-29 00:08:17 +0000
6	2017-07-28 16:27:12 +0000
7	2017-07-28 00:22:40 +0000
8	2017-07-27 16:25:51 +0000
9	2017-07-26 15:59:51 +0000
10	2017-07-26 00:31:25 +0000
11	2017-07-25 16:11:53 +0000
12	2017-07-25 01:55:32 +0000
13	2017-07-25 00:10:02 +0000
14	2017-07-24 17:02:04 +0000
15	2017-07-24 00:19:32 +0000
16	2017-07-23 00:22:39 +0000
17	2017-07-22 16:56:37 +0000
18	2017-07-22 00:23:06 +0000
19	2017-07-21 01:02:36 +0000
20	2017-07-20 16:49:33 +0000
21	2017-07-19 16:06:48 +0000
22	2017-07-19 03:39:09 +0000
23	2017-07-19 00:47:34 +0000
24	2017-07-18 16:08:03 +0000
25	2017-07-18 00:07:08 +0000
26	2017-07-17 16:17:36 +0000
27	2017-07-16 23:58:41 +0000
28	2017-07-16 20:14:00 +0000
29	2017-07-15 23:25:31 +0000
...	...
2326	2015-11-17 00:24:19 +0000
2327	2015-11-17 00:06:54 +0000
2328	2015-11-16 23:23:41 +0000
2329	2015-11-16 21:54:18 +0000
2330	2015-11-16 21:10:36 +0000
2331	2015-11-16 20:32:58 +0000
2332	2015-11-16 20:01:42 +0000
2333	2015-11-16 19:31:45 +0000
2334	2015-11-16 16:37:02 +0000
2335	2015-11-16 16:11:11 +0000
2336	2015-11-16 15:14:19 +0000

2337 2015-11-16 14:57:41 +0000
 2338 2015-11-16 04:02:55 +0000
 2339 2015-11-16 03:55:04 +0000
 2340 2015-11-16 03:44:34 +0000
 2341 2015-11-16 03:22:39 +0000
 2342 2015-11-16 02:38:37 +0000
 2343 2015-11-16 01:59:36 +0000
 2344 2015-11-16 01:52:02 +0000
 2345 2015-11-16 01:22:45 +0000
 2346 2015-11-16 01:01:59 +0000
 2347 2015-11-16 00:55:59 +0000
 2348 2015-11-16 00:49:46 +0000
 2349 2015-11-16 00:35:11 +0000
 2350 2015-11-16 00:30:50 +0000
 2351 2015-11-16 00:24:50 +0000
 2352 2015-11-16 00:04:52 +0000
 2353 2015-11-15 23:21:54 +0000
 2354 2015-11-15 23:05:30 +0000
 2355 2015-11-15 22:32:08 +0000

```

                                source \
0      <a href="http://twitter.com/download/iphone" r...
1      <a href="http://twitter.com/download/iphone" r...
2      <a href="http://twitter.com/download/iphone" r...
3      <a href="http://twitter.com/download/iphone" r...
4      <a href="http://twitter.com/download/iphone" r...
5      <a href="http://twitter.com/download/iphone" r...
6      <a href="http://twitter.com/download/iphone" r...
7      <a href="http://twitter.com/download/iphone" r...
8      <a href="http://twitter.com/download/iphone" r...
9      <a href="http://twitter.com/download/iphone" r...
10     <a href="http://twitter.com/download/iphone" r...
11     <a href="http://twitter.com/download/iphone" r...
12     <a href="http://twitter.com/download/iphone" r...
13     <a href="http://twitter.com/download/iphone" r...
14     <a href="http://twitter.com/download/iphone" r...
15     <a href="http://twitter.com/download/iphone" r...
16     <a href="http://twitter.com/download/iphone" r...
17     <a href="http://twitter.com/download/iphone" r...
18     <a href="http://twitter.com/download/iphone" r...
19     <a href="http://twitter.com/download/iphone" r...
20     <a href="http://twitter.com/download/iphone" r...
21     <a href="http://twitter.com/download/iphone" r...
22     <a href="http://twitter.com/download/iphone" r...
23     <a href="http://twitter.com/download/iphone" r...
24     <a href="http://twitter.com/download/iphone" r...
25     <a href="http://twitter.com/download/iphone" r...
26     <a href="http://twitter.com/download/iphone" r...
  
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27 <a href="http://twitter.com/download/iphone" r...
 28 <a href="http://twitter.com/download/iphone" r...
 29 <a href="http://twitter.com/download/iphone" r...
 ...
 2326 <a href="http://twitter.com/download/iphone" r...
 2327 <a href="http://twitter.com/download/iphone" r...
 2328 <a href="http://twitter.com/download/iphone" r...
 2329 <a href="http://twitter.com/download/iphone" r...
 2330 <a href="http://twitter.com/download/iphone" r...
 2331 <a href="http://twitter.com/download/iphone" r...
 2332 <a href="http://twitter.com/download/iphone" r...
 2333 <a href="http://twitter.com/download/iphone" r...
 2334 <a href="http://twitter.com/download/iphone" r...
 2335 <a href="http://twitter.com/download/iphone" r...
 2336 <a href="http://twitter.com/download/iphone" r...
 2337 <a href="http://twitter.com/download/iphone" r...
 2338 <a href="http://twitter.com/download/iphone" r...
 2339 <a href="http://twitter.com/download/iphone" r...
 2340 <a href="http://twitter.com/download/iphone" r...
 2341 <a href="http://twitter.com/download/iphone" r...
 2342 <a href="http://twitter.com/download/iphone" r...
 2343 <a href="http://twitter.com/download/iphone" r...
 2344 <a href="http://twitter.com/download/iphone" r...
 2345 <a href="http://twitter.com/download/iphone" r...
 2346 <a href="http://twitter.com/download/iphone" r...
 2347 <a href="http://twitter.com/download/iphone" r...
 2348 <a href="http://twitter.com/download/iphone" r...
 2349 <a href="http://twitter.com/download/iphone" r...
 2350 <a href="http://twitter.com/download/iphone" r...
 2351 <a href="http://twitter.com/download/iphone" r...
 2352 <a href="http://twitter.com/download/iphone" r...
 2353 <a href="http://twitter.com/download/iphone" r...
 2354 <a href="http://twitter.com/download/iphone" r...
 2355 <a href="http://twitter.com/download/iphone" r...

	text	retweeted_status_id \
0	This is Phineas. He's a mystical boy. Only eve...	NaN
1	This is Tilly. She's just checking pup on you...	NaN
2	This is Archie. He is a rare Norwegian Pouncin...	NaN
3	This is Darla. She commenced a snooze mid meal...	NaN
4	This is Franklin. He would like you to stop ca...	NaN
5	Here we have a majestic great white breaching ...	NaN
6	Meet Jax. He enjoys ice cream so much he gets ...	NaN
7	When you watch your owner call another dog a g...	NaN
8	This is Zoey. She doesn't want to be one of th...	NaN
9	This is Cassie. She is a college pup. Studying...	NaN
10	This is Koda. He is a South Australian decksha...	NaN
11	This is Bruno. He is a service shark. Only get...	NaN

12	Here's a puppo that seems to be on the fence a...	NaN
13	This is Ted. He does his best. Sometimes that'...	NaN
14	This is Stuart. He's sporting his favorite fan...	NaN
15	This is Oliver. You're witnessing one of his m...	NaN
16	This is Jim. He found a fren. Taught him how t...	NaN
17	This is Zeke. He has a new stick. Very proud o...	NaN
18	This is Ralphus. He's powering up. Attempting ...	NaN
19	RT @dog_rates: This is Canela. She attempted s...	8.874740e+17
20	This is Gerald. He was just told he didn't get...	NaN
21	This is Jeffrey. He has a monopoly on the pool...	NaN
22	I've yet to rate a Venezuelan Hover Wiener. Th...	NaN
23	This is Canela. She attempted some fancy porch...	NaN
24	You may not have known you needed to see this ...	NaN
25	This... is a Jubilant Antarctic House Bear. We...	NaN
26	This is Maya. She's very shy. Rarely leaves he...	NaN
27	This is Mingus. He's a wonderful father to his...	NaN
28	This is Derek. He's late for a dog meeting. 13...	NaN
29	This is Roscoe. Another pupper fallen victim t...	NaN
...
2326	This is quite the dog. Gets really excited whe...	NaN
2327	This is a southern Vesuvius bumblegruff. Can d...	NaN
2328	Oh goodness. A super rare northeast Qdoba kang...	NaN
2329	Those are sunglasses and a jean jacket. 11/10 ...	NaN
2330	Unique dog here. Very small. Lives in containe...	NaN
2331	Here we have a mixed Asiago from the Galápagos...	NaN
2332	Look at this jokester thinking seat belt laws ...	NaN
2333	This is an extremely rare horned Parthenon. No...	NaN
2334	This is a funny dog. Weird toes. Won't come do...	NaN
2335	This is an Albanian 3 1/2 legged Episcopalian...	NaN
2336	Can take selfies 11/10 https://t.co/ws2AMaWpPW	NaN
2337	Very concerned about fellow dog trapped in com...	NaN
2338	Not familiar with this breed. No tail (weird)...	NaN
2339	Oh my. Here you are seeing an Adobe Setter giv...	NaN
2340	Can stand on stump for what seems like a while...	NaN
2341	This appears to be a Mongolian Presbyterian mi...	NaN
2342	Here we have a well-established sunblockerspan...	NaN
2343	Let's hope this flight isn't Malaysian (lol). ...	NaN
2344	Here we have a northern speckled Rhododendron...	NaN
2345	This is the happiest dog you will ever see. Ve...	NaN
2346	Here is the Rand Paul of retrievers folks! He'...	NaN
2347	My oh my. This is a rare blond Canadian terrie...	NaN
2348	Here is a Siberian heavily armored polar bear ...	NaN
2349	This is an odd dog. Hard on the outside but lo...	NaN
2350	This is a truly beautiful English Wilson Staff...	NaN
2351	Here we have a 1949 1st generation vulpix. Enj...	NaN
2352	This is a purebred Piers Morgan. Loves to Netf...	NaN
2353	Here is a very happy pup. Big fan of well-main...	NaN
2354	This is a western brown Mitsubishi terrier. Up...	NaN

2355 Here we have a Japanese Irish Setter. Lost eye... NaN

	retweeted_status_user_id	retweeted_status_timestamp	\
0	NaN	NaN	
1	NaN	NaN	
2	NaN	NaN	
3	NaN	NaN	
4	NaN	NaN	
5	NaN	NaN	
6	NaN	NaN	
7	NaN	NaN	
8	NaN	NaN	
9	NaN	NaN	
10	NaN	NaN	
11	NaN	NaN	
12	NaN	NaN	
13	NaN	NaN	
14	NaN	NaN	
15	NaN	NaN	
16	NaN	NaN	
17	NaN	NaN	
18	NaN	NaN	
19	4.196984e+09	2017-07-19 00:47:34	+0000
20	NaN	NaN	
21	NaN	NaN	
22	NaN	NaN	
23	NaN	NaN	
24	NaN	NaN	
25	NaN	NaN	
26	NaN	NaN	
27	NaN	NaN	
28	NaN	NaN	
29	NaN	NaN	
...	
2326	NaN	NaN	
2327	NaN	NaN	
2328	NaN	NaN	
2329	NaN	NaN	
2330	NaN	NaN	
2331	NaN	NaN	
2332	NaN	NaN	
2333	NaN	NaN	
2334	NaN	NaN	
2335	NaN	NaN	
2336	NaN	NaN	
2337	NaN	NaN	
2338	NaN	NaN	
2339	NaN	NaN	

2340	NaN	NaN
2341	NaN	NaN
2342	NaN	NaN
2343	NaN	NaN
2344	NaN	NaN
2345	NaN	NaN
2346	NaN	NaN
2347	NaN	NaN
2348	NaN	NaN
2349	NaN	NaN
2350	NaN	NaN
2351	NaN	NaN
2352	NaN	NaN
2353	NaN	NaN
2354	NaN	NaN
2355	NaN	NaN

	expanded_urls	rating_numerator	\
0	https://twitter.com/dog_rates/status/892420643...	13	
1	https://twitter.com/dog_rates/status/892177421...	13	
2	https://twitter.com/dog_rates/status/891815181...	12	
3	https://twitter.com/dog_rates/status/891689557...	13	
4	https://twitter.com/dog_rates/status/891327558...	12	
5	https://twitter.com/dog_rates/status/891087950...	13	
6	https://gofundme.com/ydvmve-surgery-for-jax,ht...	13	
7	https://twitter.com/dog_rates/status/890729181...	13	
8	https://twitter.com/dog_rates/status/890609185...	13	
9	https://twitter.com/dog_rates/status/890240255...	14	
10	https://twitter.com/dog_rates/status/890006608...	13	
11	https://twitter.com/dog_rates/status/889880896...	13	
12	https://twitter.com/dog_rates/status/889665388...	13	
13	https://twitter.com/dog_rates/status/889638837...	12	
14	https://twitter.com/dog_rates/status/889531135...	13	
15	https://twitter.com/dog_rates/status/889278841...	13	
16	https://twitter.com/dog_rates/status/888917238...	12	
17	https://twitter.com/dog_rates/status/888804989...	13	
18	https://twitter.com/dog_rates/status/888554962...	13	
19	https://twitter.com/dog_rates/status/887473957...	13	
20	https://twitter.com/dog_rates/status/888078434...	12	
21	https://twitter.com/dog_rates/status/887705289...	13	
22	https://twitter.com/dog_rates/status/887517139...	14	
23	https://twitter.com/dog_rates/status/887473957...	13	
24	https://twitter.com/dog_rates/status/887343217...	13	
25	https://twitter.com/dog_rates/status/887101392...	12	
26	https://twitter.com/dog_rates/status/886983233...	13	
27	https://www.gofundme.com/mingusneedsus,https:/...	13	
28	https://twitter.com/dog_rates/status/886680336...	13	
29	https://twitter.com/dog_rates/status/886366144...	12	

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...
2326 https://twitter.com/dog_rates/status/666411507... 2
2327 https://twitter.com/dog_rates/status/666407126... 7
2328 https://twitter.com/dog_rates/status/666396247... 9
2329 https://twitter.com/dog_rates/status/666373753... 11
2330 https://twitter.com/dog_rates/status/666362758... 6
2331 https://twitter.com/dog_rates/status/666353288... 8
2332 https://twitter.com/dog_rates/status/666345417... 10
2333 https://twitter.com/dog_rates/status/666337882... 9
2334 https://twitter.com/dog_rates/status/666293911... 3
2335 https://twitter.com/dog_rates/status/666287406... 1
2336 https://twitter.com/dog_rates/status/666273097... 11
2337 https://twitter.com/dog_rates/status/666268910... 10
2338 https://twitter.com/dog_rates/status/666104133... 1
2339 https://twitter.com/dog_rates/status/666102155... 11
2340 https://twitter.com/dog_rates/status/666099513... 8
2341 https://twitter.com/dog_rates/status/666094000... 9
2342 https://twitter.com/dog_rates/status/666082916... 6
2343 https://twitter.com/dog_rates/status/666073100... 10
2344 https://twitter.com/dog_rates/status/666071193... 9
2345 https://twitter.com/dog_rates/status/666063827... 10
2346 https://twitter.com/dog_rates/status/666058600... 8
2347 https://twitter.com/dog_rates/status/666057090... 9
2348 https://twitter.com/dog_rates/status/666055525... 10
2349 https://twitter.com/dog_rates/status/666051853... 2
2350 https://twitter.com/dog_rates/status/666050758... 10
2351 https://twitter.com/dog_rates/status/666049248... 5
2352 https://twitter.com/dog_rates/status/666044226... 6
2353 https://twitter.com/dog_rates/status/666033412... 9
2354 https://twitter.com/dog_rates/status/666029285... 7
2355 https://twitter.com/dog_rates/status/666020888... 8

```

	rating_denominator	name	doggo	floofer	pupper	puppo
0	10	Phineas	None	None	None	None
1	10	Tilly	None	None	None	None
2	10	Archie	None	None	None	None
3	10	Darla	None	None	None	None
4	10	Franklin	None	None	None	None
5	10	None	None	None	None	None
6	10	Jax	None	None	None	None
7	10	None	None	None	None	None
8	10	Zoey	None	None	None	None
9	10	Cassie	doggo	None	None	None
10	10	Koda	None	None	None	None
11	10	Bruno	None	None	None	None
12	10	None	None	None	None	puppo
13	10	Ted	None	None	None	None
14	10	Stuart	None	None	None	puppo

15	10	Oliver	None	None	None	None
16	10	Jim	None	None	None	None
17	10	Zeke	None	None	None	None
18	10	Ralphus	None	None	None	None
19	10	Canela	None	None	None	None
20	10	Gerald	None	None	None	None
21	10	Jeffrey	None	None	None	None
22	10	such	None	None	None	None
23	10	Canela	None	None	None	None
24	10	None	None	None	None	None
25	10	None	None	None	None	None
26	10	Maya	None	None	None	None
27	10	Mingus	None	None	None	None
28	10	Derek	None	None	None	None
29	10	Roscoe	None	None	pupper	None
...
2326	10	quite	None	None	None	None
2327	10	a	None	None	None	None
2328	10	None	None	None	None	None
2329	10	None	None	None	None	None
2330	10	None	None	None	None	None
2331	10	None	None	None	None	None
2332	10	None	None	None	None	None
2333	10	an	None	None	None	None
2334	10	a	None	None	None	None
2335	2	an	None	None	None	None
2336	10	None	None	None	None	None
2337	10	None	None	None	None	None
2338	10	None	None	None	None	None
2339	10	None	None	None	None	None
2340	10	None	None	None	None	None
2341	10	None	None	None	None	None
2342	10	None	None	None	None	None
2343	10	None	None	None	None	None
2344	10	None	None	None	None	None
2345	10	the	None	None	None	None
2346	10	the	None	None	None	None
2347	10	a	None	None	None	None
2348	10	a	None	None	None	None
2349	10	an	None	None	None	None
2350	10	a	None	None	None	None
2351	10	None	None	None	None	None
2352	10	a	None	None	None	None
2353	10	a	None	None	None	None
2354	10	a	None	None	None	None
2355	10	None	None	None	None	None

[2356 rows x 17 columns]

```
In [14]: df_twitter.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 17 columns):
tweet_id                2356 non-null int64
in_reply_to_status_id   78 non-null float64
in_reply_to_user_id     78 non-null float64
timestamp               2356 non-null object
source                 2356 non-null object
text                   2356 non-null object
retweeted_status_id     181 non-null float64
retweeted_status_user_id 181 non-null float64
retweeted_status_timestamp 181 non-null object
expanded_urls          2297 non-null object
rating_numerator        2356 non-null int64
rating_denominator      2356 non-null int64
name                   2356 non-null object
doggo                  2356 non-null object
floofer                2356 non-null object
pupper                2356 non-null object
puppo                  2356 non-null object
dtypes: float64(4), int64(3), object(10)
memory usage: 313.0+ KB
```

```
In [15]: df_twitter.describe()
```

```
Out[15]:
```

	tweet_id	in_reply_to_status_id	in_reply_to_user_id	\
count	2.356000e+03	7.800000e+01	7.800000e+01	
mean	7.427716e+17	7.455079e+17	2.014171e+16	
std	6.856705e+16	7.582492e+16	1.252797e+17	
min	6.660209e+17	6.658147e+17	1.185634e+07	
25%	6.783989e+17	6.757419e+17	3.086374e+08	
50%	7.196279e+17	7.038708e+17	4.196984e+09	
75%	7.993373e+17	8.257804e+17	4.196984e+09	
max	8.924206e+17	8.862664e+17	8.405479e+17	

	retweeted_status_id	retweeted_status_user_id	rating_numerator	\
count	1.810000e+02	1.810000e+02	2356.000000	
mean	7.720400e+17	1.241698e+16	13.126486	
std	6.236928e+16	9.599254e+16	45.876648	
min	6.661041e+17	7.832140e+05	0.000000	
25%	7.186315e+17	4.196984e+09	10.000000	
50%	7.804657e+17	4.196984e+09	11.000000	
75%	8.203146e+17	4.196984e+09	12.000000	
max	8.874740e+17	7.874618e+17	1776.000000	

	rating_denominator
count	2356.000000
mean	10.455433
std	6.745237
min	0.000000
25%	10.000000
50%	10.000000
75%	10.000000
max	170.000000

In [16]: df_image.head()

```
Out[16]:
```

	tweet_id	jpg_url	
0	666020888022790149	https://pbs.twimg.com/media/CT4udn0WwAA0aMy.jpg	
1	666029285002620928	https://pbs.twimg.com/media/CT42GRgUYAA5iDo.jpg	
2	666033412701032449	https://pbs.twimg.com/media/CT4521TWwAEvMyu.jpg	
3	666044226329800704	https://pbs.twimg.com/media/CT5Dr8HUEAA-lEu.jpg	
4	666049248165822465	https://pbs.twimg.com/media/CT5IQmsXIAAKY4A.jpg	

	img_num	p1	p1_conf	p1_dog	p2
0	1	Welsh_springer_spaniel	0.465074	True	collie
1	1	redbone	0.506826	True	miniature_pinscher
2	1	German_shepherd	0.596461	True	malinois
3	1	Rhodesian_ridgeback	0.408143	True	redbone
4	1	miniature_pinscher	0.560311	True	Rottweiler

	p2_conf	p2_dog	p3	p3_conf	p3_dog
0	0.156665	True	Shetland_sheepdog	0.061428	True
1	0.074192	True	Rhodesian_ridgeback	0.072010	True
2	0.138584	True	bloodhound	0.116197	True
3	0.360687	True	miniature_pinscher	0.222752	True
4	0.243682	True	Doberman	0.154629	True

In [17]: df_image.tail()

```
Out[17]:
```

	tweet_id	jpg_url	
2070	891327558926688256	https://pbs.twimg.com/media/DF6hr6BUMAAzZgT.jpg	
2071	891689557279858688	https://pbs.twimg.com/media/DF_q7IAWsAEuuN8.jpg	
2072	891815181378084864	https://pbs.twimg.com/media/DGBdLU1WsAANxJ9.jpg	
2073	892177421306343426	https://pbs.twimg.com/media/DGGmoV4XsAAUL6n.jpg	
2074	892420643555336193	https://pbs.twimg.com/media/DGKD1-bXoAAIAUK.jpg	

	img_num	p1	p1_conf	p1_dog	p2	p2_conf
2070	2	basset	0.555712	True	English_springer	0.225770
2071	1	paper_towel	0.170278	False	Labrador_retriever	0.168086
2072	1	Chihuahua	0.716012	True	malamute	0.078253
2073	1	Chihuahua	0.323581	True	Pekinese	0.090647
2074	1	orange	0.097049	False	bagel	0.085851

	p2_dog		p3	p3_conf	p3_dog
2070	True	German_short-haired_pointer		0.175219	True
2071	True		spatula	0.040836	False
2072	True		kelpie	0.031379	True
2073	True		papillon	0.068957	True
2074	False		banana	0.076110	False

```
In [18]: df_image.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2075 entries, 0 to 2074
Data columns (total 12 columns):
tweet_id      2075 non-null int64
jpg_url       2075 non-null object
img_num       2075 non-null int64
p1            2075 non-null object
p1_conf       2075 non-null float64
p1_dog        2075 non-null bool
p2            2075 non-null object
p2_conf       2075 non-null float64
p2_dog        2075 non-null bool
p3            2075 non-null object
p3_conf       2075 non-null float64
p3_dog        2075 non-null bool
dtypes: bool(3), float64(3), int64(2), object(4)
memory usage: 152.1+ KB
```

```
In [19]: df_image.describe()
```

```
Out[19]:
```

	tweet_id	img_num	p1_conf	p2_conf	p3_conf
count	2.075000e+03	2075.000000	2075.000000	2.075000e+03	2.075000e+03
mean	7.384514e+17	1.203855	0.594548	1.345886e-01	6.032417e-02
std	6.785203e+16	0.561875	0.271174	1.006657e-01	5.090593e-02
min	6.660209e+17	1.000000	0.044333	1.011300e-08	1.740170e-10
25%	6.764835e+17	1.000000	0.364412	5.388625e-02	1.622240e-02
50%	7.119988e+17	1.000000	0.588230	1.181810e-01	4.944380e-02
75%	7.932034e+17	1.000000	0.843855	1.955655e-01	9.180755e-02
max	8.924206e+17	4.000000	1.000000	4.880140e-01	2.734190e-01

```
In [11]: json_tweets.head()
```

```
Out[11]:
```

	tweet_id	favorites	retweets	user_followers	user_favourites \
0	892420643555336193	38690	8558	6992402	134549
1	892177421306343426	33159	6291	6992402	134549
2	891815181378084864	24962	4174	6992402	134549
3	891689557279858688	42071	8690	6992402	134549
4	891327558926688256	40227	9450	6992402	134549

	date_time
0	2017-08-01 16:23:56
1	2017-08-01 00:17:27
2	2017-07-31 00:18:03
3	2017-07-30 15:58:51
4	2017-07-29 16:00:24

In [12]: json_tweets.tail()

Out[12]:

	tweet_id	favorites	retweets	user_followers	\
2351	666020888022790149	2563	516	6992426	
2352	666020888022790149	2563	516	6992426	
2353	666020888022790149	2563	516	6992426	
2354	666020888022790149	2563	516	6992426	
2355	666020888022790149	2563	516	6992426	

	user_favourites	date_time
2351	134549	2015-11-15 22:32:08
2352	134549	2015-11-15 22:32:08
2353	134549	2015-11-15 22:32:08
2354	134549	2015-11-15 22:32:08
2355	134549	2015-11-15 22:32:08

In [13]: json_tweets.info()

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2356 entries, 0 to 2355
Data columns (total 6 columns):
tweet_id      2356 non-null int64
favorites     2356 non-null int64
retweets      2356 non-null int64
user_followers 2356 non-null int64
user_favourites 2356 non-null int64
date_time     2356 non-null object
dtypes: int64(5), object(1)
memory usage: 110.5+ KB
```

In [14]: json_tweets.describe()

Out[14]:

	tweet_id	favorites	retweets	user_followers	\
count	2.356000e+03	2356.000000	2356.000000	2.356000e+03	
mean	7.419379e+17	8018.380730	3003.025891	6.992430e+06	
std	6.837471e+16	12081.792098	5005.958273	1.140488e+02	
min	6.660209e+17	0.000000	0.000000	6.992402e+06	
25%	6.779127e+17	1404.000000	596.750000	6.992410e+06	
50%	7.183508e+17	3512.500000	1399.000000	6.992419e+06	
75%	7.986836e+17	9911.250000	3500.500000	6.992425e+06	
max	8.924206e+17	142980.000000	77127.000000	6.993395e+06	

	user_favourites
count	2356.000000
mean	134549.014007
std	0.117543
min	134549.000000
25%	134549.000000
50%	134549.000000
75%	134549.000000
max	134550.000000

3 Clean

3.0.1 Quality

df_twitter

- in_reply_to_status_id, in_reply_to_user_id, retweeted_status_id, retweeted_status_user_id should be integers and not float
- In several columns null objects are not null
- Name column have invalid names
- We want originals tweets having ratings not retweets

df_image

- Missing values from image dataset
- Some tweet_ids have the same jpg_url
- Some tweets are have 2 different tweet_id one redirect to the other

json_tweets

- duplicated records

3.0.2 Tidiness

- Unnecessary details
- Various stages of dogs in columns instead of rows archives dataset

```
In [7]: df_master = pd.merge(df_twitter, df_image)
```

```
In [8]: df_master = pd.merge(df_master, json_tweets)
```

```
In [7]: df_master.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 2080 entries, 0 to 2079
Data columns (total 33 columns):
tweet_id                2080 non-null int64
in_reply_to_status_id    23 non-null float64
```

```

in_reply_to_user_id      23 non-null float64
timestamp                2080 non-null object
source                  2080 non-null object
text                    2080 non-null object
retweeted_status_id      75 non-null float64
retweeted_status_user_id 75 non-null float64
retweeted_status_timestamp 75 non-null object
expanded_urls            2080 non-null object
rating_numerator         2080 non-null int64
rating_denominator       2080 non-null int64
name                    2080 non-null object
doggo                   2080 non-null object
floofer                 2080 non-null object
pupper                  2080 non-null object
puppo                   2080 non-null object
jpg_url                 2080 non-null object
img_num                 2080 non-null int64
p1                      2080 non-null object
p1_conf                 2080 non-null float64
p1_dog                  2080 non-null bool
p2                      2080 non-null object
p2_conf                 2080 non-null float64
p2_dog                  2080 non-null bool
p3                      2080 non-null object
p3_conf                 2080 non-null float64
p3_dog                  2080 non-null bool
favorites                2080 non-null int64
retweets                2080 non-null int64
user_followers           2080 non-null int64
user_favourites          2080 non-null int64
date_time                2080 non-null object
dtypes: bool(3), float64(7), int64(8), object(15)
memory usage: 509.8+ KB

```

```

In [9]: df_master = df_master[pd.isnull(df_master.retweeted_status_id)]
        df_master = df_master.drop_duplicates()
        df_master = df_master.dropna(subset = ['jpg_url'])

```

```

In [10]: df_master = df_master.drop('retweeted_status_id', 1)

```

```

In [11]: df_master = df_master.drop('retweeted_status_user_id', 1)
        df_master = df_master.drop('retweeted_status_timestamp', 1)

```

```

In [12]: df_master = df_master.drop('date_time', 1)

```

```

In [12]: df_master.info()

```

```

<class 'pandas.core.frame.DataFrame'>
Int64Index: 1994 entries, 0 to 2068

```

```

Data columns (total 29 columns):
tweet_id          1994 non-null int64
in_reply_to_status_id  23 non-null float64
in_reply_to_user_id  23 non-null float64
timestamp         1994 non-null object
source            1994 non-null object
text              1994 non-null object
expanded_urls     1994 non-null object
rating_numerator   1994 non-null int64
rating_denominator 1994 non-null int64
name              1994 non-null object
doggo             1994 non-null object
floofer           1994 non-null object
pupper           1994 non-null object
puppo            1994 non-null object
jpg_url           1994 non-null object
img_num           1994 non-null int64
p1                1994 non-null object
p1_conf           1994 non-null float64
p1_dog            1994 non-null bool
p2                1994 non-null object
p2_conf           1994 non-null float64
p2_dog            1994 non-null bool
p3                1994 non-null object
p3_conf           1994 non-null float64
p3_dog            1994 non-null bool
favorites          1994 non-null int64
retweets          1994 non-null int64
user_followers    1994 non-null int64
user_favourites   1994 non-null int64
dtypes: bool(3), float64(5), int64(8), object(13)
memory usage: 426.5+ KB

```

```

In [13]: print(df_master.doggo.value_counts())
          print(df_master.floofer.value_counts())
          print(df_master.pupper.value_counts())
          print(df_master.puppo.value_counts())

```

```

None      1920
doggo      74
Name: doggo, dtype: int64
None      1986
floofer     8
Name: floofer, dtype: int64
None      1782
pupper     212
Name: pupper, dtype: int64

```

```
None      1971
puppo      23
Name: puppo, dtype: int64
```

```
In [14]: melt_columns = ['doggo', 'floofer', 'pupper', 'puppo']
        not_to_melt_columns = [x for x in df_master.columns.tolist() if x not in melt_columns]
        df_master = pd.melt(df_master, id_vars = not_to_melt_columns, value_vars = melt_columns,
                           var_name = 'stages', value_name = 'dog_stage')
```

```
In [15]: df_master = df_master.drop('stages', 1)
```

```
In [14]: print(df_master.dog_stage.value_counts())
```

```
None      7659
pupper     212
doggo       74
puppo       23
floofer      8
Name: dog_stage, dtype: int64
```

```
In [16]: df_master = df_master.sort_values('dog_stage').drop_duplicates('tweet_id', keep = 'last')
```

```
In [16]: print(df_master.dog_stage.value_counts())
```

```
None      1688
pupper     212
doggo       63
puppo       23
floofer      8
Name: dog_stage, dtype: int64
```

```
In [17]: df_master.apply(lambda x : len(x.unique()))
```

```
Out[17]: tweet_id      1994
         in_reply_to_status_id    23
         in_reply_to_user_id      2
         timestamp      1994
         source            3
         text             1994
         expanded_urls     1994
         rating_numerator    34
         rating_denominator   15
         name              936
         jpg_url           1994
         img_num            4
         p1                374
```

p1_conf	1991
p1_dog	2
p2	401
p2_conf	1989
p2_dog	2
p3	405
p3_conf	1991
p3_dog	2
favorites	1837
retweets	1600
user_followers	33
user_favourites	2
dog_stage	5
dtype: int64	

```
In [39]: df_master.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 1994 entries, 1918 to 6764
Data columns (total 26 columns):
tweet_id          1994 non-null int64
in_reply_to_status_id  23 non-null float64
in_reply_to_user_id  23 non-null float64
timestamp         1994 non-null object
source            1994 non-null object
text              1994 non-null object
expanded_urls      1994 non-null object
rating_numerator    1994 non-null int64
rating_denominator  1994 non-null int64
name              1994 non-null object
jpg_url            1994 non-null object
img_num           1994 non-null int64
p1                1994 non-null object
p1_conf            1994 non-null float64
p1_dog             1994 non-null bool
p2                1994 non-null object
p2_conf            1994 non-null float64
p2_dog             1994 non-null bool
p3                1994 non-null object
p3_conf            1994 non-null float64
p3_dog             1994 non-null bool
favorites          1994 non-null int64
retweets           1994 non-null int64
user_followers      1994 non-null int64
user_favourites     1994 non-null int64
dog_stage          1994 non-null object
dtypes: bool(3), float64(5), int64(8), object(10)
memory usage: 379.7+ KB
```

```
Out[39]: 134549      1967
         134550        27
         Name: user_favourites, dtype: int64
```

```
In [18]: df_master['in_reply_to_user_id'].value_counts()
         df_master['source'].value_counts()
         df_master['user_favourites'].value_counts()
```

```
Out[18]: 134549      1967
         134550        27
         Name: user_favourites, dtype: int64
```

```
In [19]: df_master = df_master.drop(['in_reply_to_status_id', 'in_reply_to_user_id', 'user_favou
```

```
In [20]: df_master['source'] = df_master['source'].apply(lambda x: re.findall(r'>(.*?)<', x)[0])
```

```
In [21]: df_master
```

```
Out[21]:
```

	tweet_id	timestamp	source \
1918	667405339315146752	2015-11-19 18:13:27 +0000	Twitter for iPhone
1917	667435689202614272	2015-11-19 20:14:03 +0000	Twitter for iPhone
1916	667437278097252352	2015-11-19 20:20:22 +0000	Twitter for iPhone
1915	667443425659232256	2015-11-19 20:44:47 +0000	Twitter for iPhone
1914	667453023279554560	2015-11-19 21:22:56 +0000	Twitter Web Client
1913	667455448082227200	2015-11-19 21:32:34 +0000	Twitter Web Client
1912	667470559035432960	2015-11-19 22:32:36 +0000	Twitter Web Client
1911	667491009379606528	2015-11-19 23:53:52 +0000	Twitter Web Client
1910	667495797102141441	2015-11-20 00:12:54 +0000	Twitter Web Client
1909	667502640335572993	2015-11-20 00:40:05 +0000	Twitter Web Client
1908	667509364010450944	2015-11-20 01:06:48 +0000	Twitter Web Client
1907	667517642048163840	2015-11-20 01:39:42 +0000	Twitter Web Client
1906	667524857454854144	2015-11-20 02:08:22 +0000	Twitter Web Client
1905	667530908589760512	2015-11-20 02:32:25 +0000	Twitter Web Client
1904	667534815156183040	2015-11-20 02:47:56 +0000	Twitter Web Client
1903	667538891197542400	2015-11-20 03:04:08 +0000	Twitter Web Client
1902	667544320556335104	2015-11-20 03:25:43 +0000	Twitter Web Client
1901	667546741521195010	2015-11-20 03:35:20 +0000	Twitter Web Client
1900	667549055577362432	2015-11-20 03:44:31 +0000	Twitter Web Client
1899	667724302356258817	2015-11-20 15:20:54 +0000	Twitter Web Client
1898	667728196545200128	2015-11-20 15:36:22 +0000	Twitter Web Client
1897	667766675769573376	2015-11-20 18:09:16 +0000	Twitter Web Client
1896	667773195014021121	2015-11-20 18:35:10 +0000	Twitter Web Client
1895	667782464991965184	2015-11-20 19:12:01 +0000	Twitter for iPhone
1894	667793409583771648	2015-11-20 19:55:30 +0000	Twitter for iPhone
1893	667801013445750784	2015-11-20 20:25:43 +0000	Twitter for iPhone
1892	667806454573760512	2015-11-20 20:47:20 +0000	Twitter for iPhone
1919	667393430834667520	2015-11-19 17:26:08 +0000	Twitter for iPhone
1891	667832474953625600	2015-11-20 22:30:44 +0000	Twitter for iPhone
1920	667369227918143488	2015-11-19 15:49:57 +0000	Twitter for iPhone


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...
5200 690015576308211712 2016-01-21 03:38:27 +0000 Twitter for iPhone
5513 675113801096802304 2015-12-11 00:44:07 +0000 Twitter for iPhone
5204 689905486972461056 2016-01-20 20:21:00 +0000 Twitter for iPhone
4064 874296783580663808 2017-06-12 16:06:11 +0000 Twitter for iPhone
5543 674447403907457024 2015-12-09 04:36:06 +0000 Twitter for iPhone
5057 702598099714314240 2016-02-24 20:56:55 +0000 Twitter for iPhone
5209 689623661272240129 2016-01-20 01:41:08 +0000 Twitter for iPhone
6281 825026590719483904 2017-01-27 17:04:02 +0000 Twitter for iPhone
5996 889531135344209921 2017-07-24 17:02:04 +0000 Twitter for iPhone
6680 752519690950500352 2016-07-11 15:07:30 +0000 Twitter for iPhone
6130 855851453814013952 2017-04-22 18:31:02 +0000 Twitter for iPhone
6291 822872901745569793 2017-01-21 18:26:02 +0000 Twitter for iPhone
6691 751132876104687617 2016-07-07 19:16:47 +0000 Twitter for iPhone
6448 793195938047070209 2016-10-31 21:00:23 +0000 Twitter for iPhone
6060 874012996292530176 2017-06-11 21:18:31 +0000 Twitter for iPhone
5994 889665388333682689 2017-07-25 01:55:32 +0000 Twitter for iPhone
6116 859607811541651456 2017-05-03 03:17:27 +0000 Twitter for iPhone
6481 787717603741622272 2016-10-16 18:11:26 +0000 Twitter for iPhone
6308 819952236453363712 2017-01-13 17:00:21 +0000 Twitter for iPhone
6795 738537504001953792 2016-06-03 01:07:16 +0000 Twitter for iPhone
6753 744995568523612160 2016-06-20 20:49:19 +0000 Twitter for iPhone
6661 756275833623502848 2016-07-21 23:53:04 +0000 Twitter for iPhone
6516 780931614150983680 2016-09-28 00:46:20 +0000 Twitter for iPhone
6402 802239329049477120 2016-11-25 19:55:35 +0000 Twitter for iPhone
6394 803773340896923648 2016-11-30 01:31:12 +0000 Twitter for iPhone
6463 790946055508652032 2016-10-25 16:00:09 +0000 Twitter for iPhone
6087 867421006826221569 2017-05-24 16:44:18 +0000 Twitter for iPhone
6044 878776093423087618 2017-06-25 00:45:22 +0000 Twitter for iPhone
6279 825535076884762624 2017-01-29 02:44:34 +0000 Twitter for iPhone
6764 743253157753532416 2016-06-16 01:25:36 +0000 Twitter for iPhone

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text \
1918 This is Biden. Biden just tripped... 7/10 http...
1917 Ermergerd 12/10 https://t.co/PQni2sjPsm
1916 Never seen this breed before. Very pointy pup...
1915 Exotic dog here. Long neck. Weird paws. Obsess...
1914 Meet Cupcake. I would do unspeakable things fo...
1913 This is Reese and Twips. Reese protects Twips...
1912 This is a northern Wahoo named Kohl. He runs t...
1911 Two dogs in this one. Both are rare Jujitsu Py...
1910 This is Philippe from Soviet Russia. Commandin...
1909 Say hello to Hall and Oates. Oates is winking ...
1908 This a Norwegian Pewterschmidt named Tickles. ...
1907 This is Dook & Milo. Dook is struggling to...
1906 Another topnotch dog. His name is Big Jumpy Ra...
1905 Meet Naphaniel. He doesn't necessarily enjoy h...
1904 This is Frank (pronounced "Fronq"). Too many b...

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1903 This is a southwest Coriander named Klint. Hat...
 1902 This is Kial. Kial is either wearing a cape, w...
 1901 Here is George. George took a selfie of his ne...
 1900 Never seen dog like this. Breathes heavy. Tilt...
 1899 What a dog to start the day with. Very calm. L...
 1898 Meet Olive. He comes to spot by tree to remini...
 1897 This is Calvin. He is a Luxembourgian Mayo. Ha...
 1896 This is a rare Hungarian Pinot named Jessiga. ...
 1895 Super rare dog. Endangered (?). Thinks it's fu...
 1894 Dogs only please. Small cows and other non can...
 1893 OMIGOD 12/10 <https://t.co/SVMF4Frflw>
 1892 This is Filup. He is overcome with joy after f...
 1919 This is Fwed. He is a Canadian Asian Taylormad...
 1891 THE EYES 12/10\n\nI'm sorry. These are suppose...
 1920 Here we have a neat pup. Very white. Cool shad...
 ...
 5200 This pupper can only sleep on shoes. It's a cr...
 5513 Meet Zuzu. He just graduated college. Astute p...
 5204 Say hello to Gizmo. He's quite the pupper. Con...
 4064 This is Jed. He may be the fanciest pupper in ...
 5543 This pupper just wants a belly rub. This puppe...
 5057 This is Sansa. She's gotten too big for her ch...
 5209 This is Lucy. She's terrified of the stuffed b...
 6281 Say hello to Pablo. He's one gorgeous puppo. A...
 5996 This is Stuart. He's sporting his favorite fan...
 6680 Hopefully this puppo on a swing will help get ...
 6130 Here's a puppo participating in the #ScienceMa...
 6291 Here's a super supportive puppo participating ...
 6691 This is Cooper. He's just so damn happy. 10/10...
 6448 Say hello to Lily. She's pupset that her costu...
 6060 This is Sebastian. He can't see all the colors...
 5994 Here's a puppo that seems to be on the fence a...
 6116 Sorry for the lack of posts today. I came home...
 6481 This is Tonks. She is a service puppo. Can hea...
 6308 This is Oliver. He has dreams of being a servi...
 6795 This is Bayley. She fell asleep trying to esca...
 6753 This is Abby. She got her face stuck in a glas...
 6661 When ur older siblings get to play in the deep...
 6516 I want to finally rate this iconic puppo who t...
 6402 This is Loki. He'll do your taxes for you. Can...
 6394 This is Diogi. He fell in the pool as soon as ...
 6463 This is Betty. She's assisting with the dishes...
 6087 This is Shikha. She just watched you drop a sk...
 6044 This is Snoopy. He's a proud #PrideMonthPuppo...
 6279 Here's a very loving and accepting puppo. Appe...
 6764 This is Kilo. He cannot reach the snackum. Nif...

expanded_urls rating_numerator \

1918	https://twitter.com/dog_rates/status/667405339...	7
1917	https://twitter.com/dog_rates/status/667435689...	12
1916	https://twitter.com/dog_rates/status/667437278...	10
1915	https://twitter.com/dog_rates/status/667443425...	6
1914	https://twitter.com/dog_rates/status/667453023...	11
1913	https://twitter.com/dog_rates/status/667455448...	7
1912	https://twitter.com/dog_rates/status/667470559...	11
1911	https://twitter.com/dog_rates/status/667491009...	7
1910	https://twitter.com/dog_rates/status/667495797...	9
1909	https://twitter.com/dog_rates/status/667502640...	11
1908	https://twitter.com/dog_rates/status/667509364...	12
1907	https://twitter.com/dog_rates/status/667517642...	8
1906	https://twitter.com/dog_rates/status/667524857...	12
1905	https://twitter.com/dog_rates/status/667530908...	10
1904	https://twitter.com/dog_rates/status/667534815...	8
1903	https://twitter.com/dog_rates/status/667538891...	9
1902	https://twitter.com/dog_rates/status/667544320...	10
1901	https://twitter.com/dog_rates/status/667546741...	9
1900	https://twitter.com/dog_rates/status/667549055...	1
1899	https://twitter.com/dog_rates/status/667724302...	7
1898	https://twitter.com/dog_rates/status/667728196...	11
1897	https://twitter.com/dog_rates/status/667766675...	9
1896	https://twitter.com/dog_rates/status/667773195...	8
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1894	https://twitter.com/dog_rates/status/667793409...	8
1893	https://twitter.com/dog_rates/status/667801013...	12
1892	https://twitter.com/dog_rates/status/667806454...	10
1919	https://twitter.com/dog_rates/status/667393430...	8
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5209	https://twitter.com/dog_rates/status/689623661...	10
6281	https://www.gofundme.com/my-puppys-double-cata...	12
5996	https://twitter.com/dog_rates/status/889531135...	13
6680	https://twitter.com/dog_rates/status/752519690...	11
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6291	https://twitter.com/dog_rates/status/822872901...	13
6691	https://twitter.com/dog_rates/status/751132876...	10
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6116	https://twitter.com/dog_rates/status/859607811...	13

6481	https://twitter.com/dog_rates/status/787717603...	13
6308	https://www.gofundme.com/servicedogoliver,http...	13
6795	https://twitter.com/dog_rates/status/738537504...	11
6753	https://twitter.com/dog_rates/status/744995568...	9
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6044	https://twitter.com/dog_rates/status/878776093...	13
6279	https://twitter.com/dog_rates/status/825535076...	14
6764	https://twitter.com/dog_rates/status/743253157...	10

	rating_denominator	name \
1918	10	Biden
1917	10	None
1916	10	None
1915	10	None
1914	10	Cupcake
1913	10	Reese
1912	10	a
1911	10	None
1910	10	Philippe
1909	10	Hall
1908	10	None
1907	10	Dook
1906	10	None
1905	10	Naphaniel
1904	10	Frank
1903	10	a
1902	10	Kial
1901	10	George
1900	10	None
1899	10	None
1898	10	Olive
1897	10	Calvin
1896	10	a
1895	10	None
1894	10	None
1893	10	None
1892	10	Filup
1919	10	Fwed
1891	10	None
1920	10	None
...
5200	10	None
5513	10	Zuzu

5204	10	Gizmo
4064	10	Jed
5543	10	None
5057	10	Sansa
5209	10	Lucy
6281	10	Pablo
5996	10	Stuart
6680	10	None
6130	10	None
6291	10	None
6691	10	Cooper
6448	10	Lily
6060	10	Sebastian
5994	10	None
6116	10	None
6481	10	Tonks
6308	10	Oliver
6795	10	Bayley
6753	10	Abby
6661	10	None
6516	10	None
6402	10	Loki
6394	10	Diogi
6463	10	Betty
6087	10	Shikha
6044	10	Snoopy
6279	10	None
6764	10	Kilo

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1917	https://pbs.twimg.com/media/CUM10HCW4AEgGSi.jpg	1	...	
1916	https://pbs.twimg.com/media/CUM2qWaWoAUZO6L.jpg	1	...	
1915	https://pbs.twimg.com/media/CUM8QZwW4AAVsBl.jpg	1	...	
1914	https://pbs.twimg.com/media/CUNE_OSUwAAAdHhX.jpg	1	...	
1913	https://pbs.twimg.com/media/CUNHMTU8AAS3HH.jpg	1	...	
1912	https://pbs.twimg.com/media/CUNU78YWEAECmpB.jpg	1	...	
1911	https://pbs.twimg.com/media/CUNniSlUYAEj1Jl.jpg	1	...	
1910	https://pbs.twimg.com/media/CUNr4-7UwAAg2lq.jpg	1	...	
1909	https://pbs.twimg.com/media/CUNyHTMUyAAQVch.jpg	1	...	
1908	https://pbs.twimg.com/media/CUN40r5UAAAAa5K4.jpg	1	...	
1907	https://pbs.twimg.com/media/CUN_wiBUkAAakTO.jpg	1	...	
1906	https://pbs.twimg.com/media/CUOGUfJW4AA_eni.jpg	1	...	
1905	https://pbs.twimg.com/media/CUOL0uGUkAAx7yh.jpg	1	...	
1904	https://pbs.twimg.com/media/CUOPYI5UcAAj_nO.jpg	1	...	
1903	https://pbs.twimg.com/media/CUOTFZOW4AABsfW.jpg	1	...	
1902	https://pbs.twimg.com/media/CUOYBbbWIAAXQGU.jpg	1	...	
1901	https://pbs.twimg.com/media/CU0aOWXWcAAO_Jy.jpg	1	...	

1900	https://pbs.twimg.com/media/CU0cVCwWsAERUKY.jpg	1	...
1899	https://pbs.twimg.com/media/CUQ7tv3W4AA3K1I.jpg	1	...
1898	https://pbs.twimg.com/media/CUQ_QahUAAAVQjn.jpg	1	...
1897	https://pbs.twimg.com/media/CURiQMnUAAAPT2M.jpg	1	...
1896	https://pbs.twimg.com/media/CURoLrOVEAAaWdR.jpg	1	...
1895	https://pbs.twimg.com/media/CURwm3cUkAARcO6.jpg	1	...
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1891	https://pbs.twimg.com/media/CUSeGFNW4AAyyHC.jpg	1	...
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5057	https://pbs.twimg.com/media/CcAhPevW8AAoknv.jpg	1	...
5209	https://pbs.twimg.com/media/CZIJ2SWIAMJgNI.jpg	1	...
6281	https://pbs.twimg.com/media/C3MVTeHWcAAGNfx.jpg	2	...
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6680	https://pbs.twimg.com/media/CnF8qVDWYAAhOg1.jpg	3	...
6130	https://pbs.twimg.com/media/C-CYWrVWAAU8AXH.jpg	1	...
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6691	https://pbs.twimg.com/media/CmyPXNOW8AEtaJ-.jpg	1	...
6448	https://pbs.twimg.com/media/CwH_foYWgAEvTyI.jpg	2	...
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5994	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg	1	...
6116	https://pbs.twimg.com/media/C-3wvtXcAUTuBE.jpg	1	...
6481	https://pbs.twimg.com/media/Cu6I9vvWIAAZGOa.jpg	3	...
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6753	https://pbs.twimg.com/media/C1bBg4WWEAMjwJu.jpg	1	...
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6044	https://pbs.twimg.com/media/DDIKMXzW0AEibje.jpg	2	...
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6764	https://pbs.twimg.com/media/C1CQzFUUYAA5vAu.jpg	1	...

	p2	p2_conf	p2_dog	p3 \
1918	Leonberg	0.127998	True	golden_retriever
1917	miniature_pinscher	0.000450	True	black-and-tan_coonhound
1916	bath_towel	0.006300	False	badger

1915	drake	0.006918	False	hen
1914	French_bulldog	0.056639	True	Staffordshire_bullterrier
1913	Irish_terrier	0.054933	True	Yorkshire_terrier
1912	pug	0.223427	True	Lakeland_terrier
1911	ice_bear	0.132264	False	weasel
1910	Christmas_stocking	0.118651	False	ski_mask
1909	golden_retriever	0.001688	True	beagle
1908	Labrador_retriever	0.119256	True	golden_retriever
1907	standard_poodle	0.084572	True	cocker_spaniel
1906	dhole	0.092435	False	Chesapeake_Bay_retriever
1905	kuvasz	0.146391	True	Labrador_retriever
1904	Cardigan	0.307407	True	cocker_spaniel
1903	silky_terrier	0.300313	True	Australian_terrier
1902	Pembroke	0.312958	True	Chihuahua
1901	miniature_poodle	0.202225	True	teddy
1900	spotlight	0.007737	False	lampshade
1899	bighorn	0.125119	False	ram
1898	golden_retriever	0.293744	True	Labrador_retriever
1897	tow_truck	0.074734	False	jeep
1896	pug	0.093494	True	ice_bear
1895	hummingbird	0.083011	False	African_grey
1894	English_setter	0.451219	True	Great_Dane
1893	Chesapeake_Bay_retriever	0.262239	True	curly-coated_retriever
1892	Chihuahua	0.187155	True	Brabancon_griffon
1919	Border_collie	0.271963	True	collie
1891	bath_towel	0.146789	False	Chihuahua
1920	bath_towel	0.127285	False	Christmas_stocking
...
5200	Siberian_husky	0.033084	True	Eskimo_dog
5513	quill	0.108807	False	joystick
5204	Shetland_sheepdog	0.023675	True	chow
4064	miniature_poodle	0.277191	True	toy_poodle
5543	malinois	0.244649	True	bull_mastiff
5057	badger	0.133584	False	Siamese_cat
5209	mashed_potato	0.208564	False	Labrador_retriever
6281	Siberian_husky	0.467678	True	malamute
5996	Labrador_retriever	0.013834	True	redbone
6680	Labrador_retriever	0.000010	True	Eskimo_dog
6130	Labrador_retriever	0.115138	True	groenendael
6291	Labrador_retriever	0.160329	True	Irish_terrier
6691	Chesapeake_Bay_retriever	0.038254	True	golden_retriever
6448	golden_retriever	0.074100	True	Chihuahua
6060	Pembroke	0.116622	True	kelpie
5994	Cardigan	0.027356	True	basenji
6116	Irish_setter	0.024099	True	Labrador_retriever
6481	malinois	0.004920	True	kelpie
6308	Staffordshire_bullterrier	0.036221	True	Italian_greyhound
6795	gibbon	0.028942	False	Pembroke

6753	Shih-Tzu	0.146336	True	Tibetan_terrier
6661	Irish_terrier	0.086981	True	bloodhound
6516	necklace	0.065462	False	chain
6402	Siberian_husky	0.335774	True	malamute
6394	redbone	0.059707	True	Irish_terrier
6463	golden_retriever	0.245773	True	chow
6087	Siberian_husky	0.381330	True	malamute
6044	whippet	0.150487	True	Ibizan_hound
6279	Tibetan_mastiff	0.147940	True	black-and-tan_coonhound
6764	Siberian_husky	0.368137	True	Eskimo_dog

	p3_conf	p3_dog	favorites	retweets	user_followers	dog_stage
1918	0.069357	True	474	224	6992426	None
1917	0.000157	True	310	85	6992426	None
1916	0.000966	False	458	248	6992426	None
1915	0.005255	False	803	604	6992426	None
1914	0.054018	True	320	93	6992426	None
1913	0.040576	True	195	65	6992426	None
1912	0.073316	True	264	100	6992426	None
1911	0.005730	False	547	235	6992426	None
1910	0.092482	False	540	286	6992426	None
1909	0.000712	True	544	227	6992426	None
1908	0.082549	True	6956	2210	6992426	None
1907	0.081347	True	377	198	6992426	None
1906	0.088122	True	1745	1162	6992426	None
1905	0.046184	True	487	253	6992426	None
1904	0.033158	True	841	559	6992426	None
1903	0.053412	True	209	69	6992426	None
1902	0.071960	True	889	547	6992426	None
1901	0.004047	False	341	130	6992426	None
1900	0.001901	False	5961	2386	6993394	None
1899	0.074673	False	502	332	6992426	None
1898	0.270673	True	383	156	6992426	None
1897	0.012773	False	456	228	6992426	None
1896	0.069038	False	240	59	6992426	None
1895	0.054247	False	423	253	6992426	None
1894	0.008164	True	718	341	6992426	None
1893	0.048920	True	338	98	6992426	None
1892	0.112799	True	1084	520	6992426	None
1919	0.073473	True	199	58	6992426	None
1891	0.104152	True	294	65	6992426	None
1920	0.028568	False	375	168	6992426	None
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5200	0.016663	True	2681	807	6992420	pupper
5513	0.043312	False	2046	839	6992426	pupper
5204	0.007165	True	2593	759	6992421	pupper
4064	0.157402	True	26058	4150	6992403	pupper
5543	0.074820	True	1103	381	6992426	pupper

5057	0.074440	False	11044	3583	6992419	pupper
5209	0.077481	True	2391	715	6992421	pupper
6281	0.004976	True	6856	1432	6992407	puppo
5996	0.007958	True	15058	2243	6992403	puppo
6680	0.000001	True	7939	3790	6992416	puppo
6130	0.096100	True	46856	18650	6992403	puppo
6291	0.069126	True	142980	48957	6992407	puppo
6691	0.007610	True	5468	1438	6992416	puppo
6448	0.042339	True	16672	6326	6992411	puppo
6060	0.049182	True	34680	10574	6992403	puppo
5994	0.004633	True	48011	10108	6992403	puppo
6116	0.019285	True	19036	1640	6992403	puppo
6481	0.000853	True	11130	3125	6992410	puppo
6308	0.020412	True	5761	1322	6992407	puppo
6795	0.026498	True	5434	1708	6992418	puppo
6753	0.134269	True	3194	692	6992417	puppo
6661	0.086276	True	6941	1697	6992416	puppo
6516	0.036469	False	23535	8237	6993378	puppo
6402	0.134589	True	9877	2927	6992412	puppo
6394	0.034195	True	10954	3108	6992411	puppo
6463	0.039012	True	18122	5291	6992411	puppo
6087	0.001670	True	16341	2583	6992403	puppo
6044	0.039725	True	19371	4155	6992403	puppo
6279	0.024525	True	56057	19124	6992407	puppo
6764	0.177822	True	4516	1326	6992417	puppo

[1994 rows x 23 columns]

```
In [22]: df_master.rating_numerator.value_counts()
df_master.rating_denominator.value_counts()
print(df_master[df_master.rating_denominator == 170]['text'][2842])
print(df_master[df_master.rating_numerator == 1776]['text'][2720])
print(df_master[df_master.tweet_id == 786709082849828864]['text'][2497])
print(df_master['text'][1918])
print(df_master['text'][1917])
print(df_master['text'][1916])
print(df_master['text'][1911])
```

Say hello to this unbelievably well behaved squad of doggos. 204/170 would try to pet all at once.
This is Atticus. He's quite simply America af. 1776/10 <https://t.co/GRXwMxLBkh>
This is Logan, the Chow who lived. He solemnly swears he's up to lots of good. H*ckin magical af.
This is Biden. Biden just tripped... 7/10 <https://t.co/3Fm9PwLju1>
Ermergerd 12/10 <https://t.co/PQni2sjPsm>
Never seen this breed before. Very pointy pup. Hurts when you cuddle. Still cute tho. 10/10 <https://t.co/3Fm9PwLju1>
Two dogs in this one. Both are rare Jujitsu Pythagoreans. One slightly whiter than other. Long 1

```
In [38]: ratings = df_master['text'].apply(lambda x: re.findall(r'(\d+(\.\d+)|(\d+))\s*/(\d+0)', x))
```

```

In [42]: rating_numerator = []
         rating_denominator = []
         dogs_count = []

In [48]: for rate in ratings:

         if len(rate) == 0:
             rating_numerator.append('NaN')
             rating_denominator.append('NaN')
             dogs_count.append(1)

         elif len(rate) == 1:
             rating_numerator.append((float(rate[0][0]) / (float(rate[0][-1])/10)))
             rating_denominator.append(float(rate[0][-1]))
             dogs_count.append(float(rate[0][-1]) / 10)

         elif len(rate) > 1 and rate[0][-1] == '10':
             rating_plus = 0
             rating_avg = 0
             for i in range(len(rate)):
                 rating_plus = rating_plus + float(rate[i][0])
             result_avg = (rating_plus / len(rate))
             rating_numerator.append(result_avg)
             rating_denominator.append(10)
             dogs_count.append(len(rate))
         else:
             rating_numerator.append('Error')
             rating_denominator.append('Error')
             dogs_count.append('Error')

df_master['new_rating_numerator'] = rating_numerator
df_master['new_rating_denominator'] = rating_denominator
df_master['dogs_count'] = dogs_count
df_master['new_rating_numerator'].value_counts()

```

```

Out[48]: 12.0          453
         10.0          411
         11.0          399
         13.0          261
         9.0           152
         8.0           94
         7.0           52
         14.0           36
         6.0           32
         5.0           30
         3.0           19
         4.0           14

```

2.0	9
1.0	4
8.5	4
9.5	3
7.5	3
10.5	2
Error	2
0.0	2
1776.0	1
9.75	1
4.5	1
5.5	1
6.5	1
NaN	1
11.26	1
13.5	1
9.666666666666666	1
420.0	1
11.5	1
11.27	1

Name: new_rating_numerator, dtype: int64

```
In [49]: print(df_master[df_master.new_rating_numerator == 'Error']['text'][2919])
print(df_master[df_master.new_rating_numerator == 'Error']['text'][2885])
print(df_master[df_master.new_rating_numerator == 1776.0]['text'][2720])
print(df_master[df_master.new_rating_numerator == 420.0]['text'][3712])
```

This is Bluebert. He just saw that both #FinalFur match ups are split 50/50. Amazed af. 11/10 ht
 Happy 4/20 from the squad! 13/10 for all <https://t.co/eV1diwds8a>
 This is Atticus. He's quite simply America af. 1776/10 <https://t.co/GRXwMxLBkh>
 After so many requests... here you go.

Good dogg. 420/10 <https://t.co/yfAAo1gdeY>

```
In [50]: tweet_id_11 = df_master[df_master.new_rating_numerator == 'Error']['tweet_id'][2919]
tweet_id_13 = df_master[df_master.new_rating_numerator == 'Error']['tweet_id'][2885]

df_master.loc[df_master['tweet_id'] == tweet_id_11, 'new_rating_numerator'] = 11
df_master.loc[df_master['tweet_id'] == tweet_id_13, 'new_rating_numerator'] = 13

df_master.loc[df_master['dogs_count'] == 'Error', 'dogs_count'] = 1
df_master.loc[df_master['new_rating_denominator'] == 'Error', 'new_rating_denominator']

In [51]: print(df_master.new_rating_numerator[df_master.tweet_id == tweet_id_11])
print(df_master.new_rating_numerator[df_master.tweet_id == tweet_id_13])
```

```
2919    11
Name: new_rating_numerator, dtype: object
```

2885 13

Name: new_rating_numerator, dtype: object

```
In [52]: df_master = df_master.drop(['rating_numerator', 'rating_denominator'], 1)
```

```
# Rename columns
```

```
df_master.rename(columns = {'new_rating_numerator': 'rating_numerator',  
                             'new_rating_denominator': 'rating_denominator'}, inplace =
```

```
list(df_master)
```

```
Out[52]: ['tweet_id',  
          'timestamp',  
          'source',  
          'text',  
          'expanded_urls',  
          'name',  
          'jpg_url',  
          'img_num',  
          'p1',  
          'p1_conf',  
          'p1_dog',  
          'p2',  
          'p2_conf',  
          'p2_dog',  
          'p3',  
          'p3_conf',  
          'p3_dog',  
          'favorites',  
          'retweets',  
          'user_followers',  
          'dog_stage',  
          'rating_numerator',  
          'rating_denominator',  
          'dogs_count']
```

```
In [53]: dog_names = []
```

```
for text in df_master['text']:
```

```
    if text.startswith('This is ') and re.match(r'[A-Z].*', text.split()[2]):  
        dog_names.append(text.split()[2].strip(',').strip('.'))
```

```
    elif text.startswith('Meet ') and re.match(r'[A-Z].*', text.split()[1]):  
        dog_names.append(text.split()[1].strip(',').strip('.'))
```

```

elif text.startswith('Say hello to ') and re.match(r'[A-Z].*', text.split()[3]):
    dog_names.append(text.split()[3].strip(',').strip('.'))

elif text.startswith('Here we have ') and re.match(r'[A-Z].*', text.split()[3]):
    dog_names.append(text.split()[3].strip(',').strip('.'))

elif 'named' in text and re.match(r'[A-Z].*', text.split()[text.split().index('named') + 1].strip(',').strip('.')):
    dog_names.append(text.split()[text.split().index('named') + 1].strip(',').strip('.'))

else:
    dog_names.append('NaN')

# Test
len(dog_names)

df_master['dog_name'] = dog_names

print("New column dog_name count \n", df_master.dog_name.value_counts())
print("Old column name count \n", df_master.name.value_counts())

```

New column dog_name count

NaN	625
Charlie	11
Cooper	10
Lucy	10
Oliver	10
Penny	9
Tucker	9
Winston	8
Sadie	8
Daisy	7
Lola	7
Jax	6
Toby	6
Bella	6
Stanley	6
Koda	6
Bo	6
Chester	5
Scout	5
Rusty	5
Louis	5
Milo	5
Oscar	5
Buddy	5
Bailey	5
Leo	5
Larry	4

Jack	4
Clarence	4
Archie	4

...

Karma	1
Ben	1
Gerbald	1
Derby	1
Glenn	1
Beemo	1
Ralphson	1
Jay	1
Herb	1
Damon	1
Pluto	1
Scott	1
Aubie	1
Jockson	1
Bruiser	1
Canela	1
Cermet	1
Ronnie	1
Kreggory	1
Marlee	1
Jacob	1
Reptar	1
Iggy	1
Tayzie	1
Pancho	1
Sully	1
Aja	1
Franq	1
Alf	1
Huxley	1

Name: dog_name, Length: 934, dtype: int64

Old column name count

None	546
a	55
Charlie	11
Cooper	10
Lucy	10
Oliver	10
Penny	9
Tucker	9
Sadie	8
Winston	8
Toby	7
the	7

Lola	7
Daisy	7
Koda	6
an	6
Bella	6
Jax	6
Stanley	6
Bo	6
Buddy	5
Bailey	5
Louis	5
Oscar	5
Leo	5
Milo	5
Rusty	5
Chester	5
Scout	5
Dave	5
...	
Alexanderson	1
Ace	1
Bert	1
Dido	1
Rose	1
Strider	1
Aubie	1
Craig	1
Jockson	1
Bruiser	1
Canela	1
Cermet	1
Marlee	1
Reptar	1
Iggy	1
Tayzie	1
Sully	1
Aja	1
Franq	1
Alf	1
Jay	1
Beemo	1
Rodney	1
Glenn	1
Tater	1
Kial	1
Eve	1
Lenox	1
Glacier	1

Huxley 1
Name: name, Length: 936, dtype: int64

```
In [54]: df_master[df_master.dog_name == 'NaN']
```

```
Out [54]:
```

	tweet_id	timestamp	source \
1917	667435689202614272	2015-11-19 20:14:03 +0000	Twitter for iPhone
1916	667437278097252352	2015-11-19 20:20:22 +0000	Twitter for iPhone
1915	667443425659232256	2015-11-19 20:44:47 +0000	Twitter for iPhone
1911	667491009379606528	2015-11-19 23:53:52 +0000	Twitter Web Client
1906	667524857454854144	2015-11-20 02:08:22 +0000	Twitter Web Client
1901	667546741521195010	2015-11-20 03:35:20 +0000	Twitter Web Client
1900	667549055577362432	2015-11-20 03:44:31 +0000	Twitter Web Client
1899	667724302356258817	2015-11-20 15:20:54 +0000	Twitter Web Client
1895	667782464991965184	2015-11-20 19:12:01 +0000	Twitter for iPhone
1894	667793409583771648	2015-11-20 19:55:30 +0000	Twitter for iPhone
1893	667801013445750784	2015-11-20 20:25:43 +0000	Twitter for iPhone
1891	667832474953625600	2015-11-20 22:30:44 +0000	Twitter for iPhone
1920	667369227918143488	2015-11-19 15:49:57 +0000	Twitter for iPhone
1948	666786068205871104	2015-11-18 01:12:41 +0000	Twitter for iPhone
1945	666826780179869698	2015-11-18 03:54:28 +0000	Twitter for iPhone
1944	666835007768551424	2015-11-18 04:27:09 +0000	Twitter for iPhone
1943	666837028449972224	2015-11-18 04:35:11 +0000	Twitter for iPhone
1939	667044094246576128	2015-11-18 18:17:59 +0000	Twitter for iPhone
1937	667065535570550784	2015-11-18 19:43:11 +0000	Twitter for iPhone
1936	667073648344346624	2015-11-18 20:15:26 +0000	Twitter for iPhone
1933	667138269671505920	2015-11-19 00:32:12 +0000	Twitter for iPhone
1927	667176164155375616	2015-11-19 03:02:47 +0000	Twitter for iPhone
1926	667177989038297088	2015-11-19 03:10:02 +0000	Twitter for iPhone
1924	667188689915760640	2015-11-19 03:52:34 +0000	Twitter for iPhone
1923	667192066997374976	2015-11-19 04:05:59 +0000	Twitter for iPhone
1888	667873844930215936	2015-11-21 01:15:07 +0000	Twitter for iPhone
1883	667911425562669056	2015-11-21 03:44:27 +0000	Twitter for iPhone
1880	667937095915278337	2015-11-21 05:26:27 +0000	Twitter for iPhone
1878	668142349051129856	2015-11-21 19:02:04 +0000	Twitter for iPhone
1877	668154635664932864	2015-11-21 19:50:53 +0000	Twitter for iPhone
...
5219	688916208532455424	2016-01-18 02:49:58 +0000	Twitter for iPhone
5051	703268521220972544	2016-02-26 17:20:56 +0000	Twitter for iPhone
5048	703407252292673536	2016-02-27 02:32:12 +0000	Twitter for iPhone
4771	743222593470234624	2016-06-15 23:24:09 +0000	Twitter for iPhone
4774	742465774154047488	2016-06-13 21:16:49 +0000	Twitter for iPhone
5488	675740360753160193	2015-12-12 18:13:51 +0000	Twitter for iPhone
5227	688519176466644993	2016-01-17 00:32:18 +0000	Twitter for iPhone
4290	824325613288833024	2017-01-25 18:38:36 +0000	Twitter for iPhone
4783	741067306818797568	2016-06-10 00:39:48 +0000	Twitter for iPhone
5477	675898130735476737	2015-12-13 04:40:46 +0000	Twitter for iPhone

5036	704761120771465216	2016-03-01 20:11:59 +0000	Twitter for iPhone
5234	687818504314159109	2016-01-15 02:08:05 +0000	Twitter for iPhone
5033	704859558691414016	2016-03-02 02:43:09 +0000	Twitter for iPhone
5468	676263575653122048	2015-12-14 04:52:55 +0000	Twitter for iPhone
4037	881536004380872706	2017-07-02 15:32:16 +0000	Twitter for iPhone
5540	674638615994089473	2015-12-09 17:15:54 +0000	Twitter for iPhone
5070	700864154249383937	2016-02-20 02:06:50 +0000	Twitter for iPhone
4249	831315979191906304	2017-02-14 01:35:49 +0000	Twitter Web Client
5505	675334060156301312	2015-12-11 15:19:21 +0000	Twitter for iPhone
5078	700151421916807169	2016-02-18 02:54:41 +0000	Twitter for iPhone
5200	690015576308211712	2016-01-21 03:38:27 +0000	Twitter for iPhone
5543	674447403907457024	2015-12-09 04:36:06 +0000	Twitter for iPhone
6680	752519690950500352	2016-07-11 15:07:30 +0000	Twitter for iPhone
6130	855851453814013952	2017-04-22 18:31:02 +0000	Twitter for iPhone
6291	822872901745569793	2017-01-21 18:26:02 +0000	Twitter for iPhone
5994	889665388333682689	2017-07-25 01:55:32 +0000	Twitter for iPhone
6116	859607811541651456	2017-05-03 03:17:27 +0000	Twitter for iPhone
6661	756275833623502848	2016-07-21 23:53:04 +0000	Twitter for iPhone
6516	780931614150983680	2016-09-28 00:46:20 +0000	Twitter for iPhone
6279	825535076884762624	2017-01-29 02:44:34 +0000	Twitter for iPhone

text \

1917 Ermergerd 12/10 <https://t.co/PQni2sjPsm>

1916 Never seen this breed before. Very pointy pup...

1915 Exotic dog here. Long neck. Weird paws. Obsess...

1911 Two dogs in this one. Both are rare Jujitsu Py...

1906 Another topnotch dog. His name is Big Jumpy Ra...

1901 Here is George. George took a selfie of his ne...

1900 Never seen dog like this. Breathes heavy. Tilt...

1899 What a dog to start the day with. Very calm. L...

1895 Super rare dog. Endangered (?). Thinks it's fu...

1894 Dogs only please. Small cows and other non can...

1893 OMIGOD 12/10 <https://t.co/SVMF4Frflw>

1891 THE EYES 12/10\n\nI'm sorry. These are suppose...

1920 Here we have a neat pup. Very white. Cool shad...

1948 Unfamiliar with this breed. Ears pointy af. Wo...

1945 12/10 simply brilliant pup <https://t.co/V6ZzG4...>

1944 These are Peruvian Feldspars. Their names are ...

1943 My goodness. Very rare dog here. Large. Tail d...

1939 12/10 gimme now <https://t.co/QZAnwgnOMB>

1937 Here we have a Hufflepuff. Loves vest. Eyes wi...

1936 Here is Dave. He is actually just a skinny leg...

1933 Extremely intelligent dog here. Has learned to...

1927 These are strange dogs. All have toupees. Long...

1926 This is a Dasani Kingfisher from Maine. His na...

1924 Quite an advanced dog here. Impressively dress...

1923 *takes several long deep breaths* omg omg oMG ...

1888 Neat dog. Lots of spikes. Always in push-up po...

1883 Wow. Armored dog here. Ready for battle. Face ...
 1880 This dog resembles a baked potato. Bed looks u...
 1878 This lil pup is Oliver. Hops around. Has wings...
 1877 Fun dogs here. Top one clearly an athlete. Bot...
 ...
 5219 This pupper just wants to say hello. 11/10 wou...
 5051 Happy Friday here's a sleepy pupper 12/10 http...
 5048 This pupper doesn't understand gates. 10/10 so...
 4771 This is a very rare Great Alaskan Bush Pupper...
 4774 Was just informed about this hero pupper and o...
 5488 Here's a pupper licking in slow motion. 12/10 ...
 5227 This pupper is sprouting a flower out of her h...
 4290 Retweet the h*ck out of this 13/10 pupper #Bel...
 4783 This is just downright precious af. 12/10 for ...
 5477 I'm sure you've all seen this pupper. Not prep...
 5036 This pupper killed this great white in an epic...
 5234 With great pupper comes great responsibility. ...
 5033 Here is a heartbreaking scene of an incredible...
 5468 All this pupper wanted to do was go skiing. No...
 4037 Here is a pupper approaching maximum borkdrive...
 5540 This pupper is fed up with being tickled. 12/1...
 5070 "Pupper is a present to world. Here is a bow f...
 4249 I couldn't make it to the #WKCDogShow BUT I ha...
 5505 Good morning here's a grass pupper. 12/10 http...
 5078 If a pupper gave that to me I'd probably start...
 5200 This pupper can only sleep on shoes. It's a cr...
 5543 This pupper just wants a belly rub. This puppe...
 6680 Hopefully this puppo on a swing will help get ...
 6130 Here's a puppo participating in the #ScienceMa...
 6291 Here's a super supportive puppo participating ...
 5994 Here's a puppo that seems to be on the fence a...
 6116 Sorry for the lack of posts today. I came home...
 6661 When ur older siblings get to play in the deep...
 6516 I want to finally rate this iconic puppo who t...
 6279 Here's a very loving and accepting puppo. Appe...

	expanded_urls	name \
1917	https://twitter.com/dog_rates/status/667435689...	None
1916	https://twitter.com/dog_rates/status/667437278...	None
1915	https://twitter.com/dog_rates/status/667443425...	None
1911	https://twitter.com/dog_rates/status/667491009...	None
1906	https://twitter.com/dog_rates/status/667524857...	None
1901	https://twitter.com/dog_rates/status/667546741...	George
1900	https://twitter.com/dog_rates/status/667549055...	None
1899	https://twitter.com/dog_rates/status/667724302...	None
1895	https://twitter.com/dog_rates/status/667782464...	None
1894	https://twitter.com/dog_rates/status/667793409...	None
1893	https://twitter.com/dog_rates/status/667801013...	None

1891	https://twitter.com/dog_rates/status/667832474...	None
1920	https://twitter.com/dog_rates/status/667369227...	None
1948	https://twitter.com/dog_rates/status/666786068...	None
1945	https://twitter.com/dog_rates/status/666826780...	None
1944	https://twitter.com/dog_rates/status/666835007...	None
1943	https://twitter.com/dog_rates/status/666837028...	None
1939	https://twitter.com/dog_rates/status/667044094...	None
1937	https://twitter.com/dog_rates/status/667065535...	None
1936	https://twitter.com/dog_rates/status/667073648...	Dave
1933	https://twitter.com/dog_rates/status/667138269...	None
1927	https://twitter.com/dog_rates/status/667176164...	None
1926	https://twitter.com/dog_rates/status/667177989...	a
1924	https://twitter.com/dog_rates/status/667188689...	None
1923	https://twitter.com/dog_rates/status/667192066...	None
1888	https://twitter.com/dog_rates/status/667873844...	None
1883	https://twitter.com/dog_rates/status/667911425...	None
1880	https://twitter.com/dog_rates/status/667937095...	None
1878	https://twitter.com/dog_rates/status/668142349...	None
1877	https://twitter.com/dog_rates/status/668154635...	None
...
5219	https://twitter.com/dog_rates/status/688916208...	None
5051	https://twitter.com/dog_rates/status/703268521...	None
5048	https://twitter.com/dog_rates/status/703407252...	None
4771	https://twitter.com/dog_rates/status/743222593...	a
4774	https://twitter.com/dog_rates/status/742465774...	None
5488	https://twitter.com/dog_rates/status/675740360...	None
5227	https://twitter.com/dog_rates/status/688519176...	None
4290	https://twitter.com/dog_rates/status/824325613...	None
4783	https://twitter.com/dog_rates/status/741067306...	just
5477	https://twitter.com/dog_rates/status/675898130...	None
5036	https://twitter.com/dog_rates/status/704761120...	None
5234	https://twitter.com/dog_rates/status/687818504...	None
5033	https://twitter.com/dog_rates/status/704859558...	a
5468	https://twitter.com/dog_rates/status/676263575...	None
4037	https://twitter.com/dog_rates/status/881536004...	a
5540	https://twitter.com/dog_rates/status/674638615...	None
5070	https://twitter.com/dog_rates/status/700864154...	a
4249	https://twitter.com/dog_rates/status/831315979...	None
5505	https://twitter.com/dog_rates/status/675334060...	None
5078	https://twitter.com/dog_rates/status/700151421...	None
5200	https://twitter.com/dog_rates/status/690015576...	None
5543	https://twitter.com/dog_rates/status/674447403...	None
6680	https://twitter.com/dog_rates/status/752519690...	None
6130	https://twitter.com/dog_rates/status/855851453...	None
6291	https://twitter.com/dog_rates/status/822872901...	None
5994	https://twitter.com/dog_rates/status/889665388...	None
6116	https://twitter.com/dog_rates/status/859607811...	None
6661	https://twitter.com/dog_rates/status/756275833...	None

6516	https://twitter.com/dog_rates/status/780931614...	None
6279	https://twitter.com/dog_rates/status/825535076...	None

	jpg_url	img_num	\
1917	https://pbs.twimg.com/media/CUM10HCW4AEgGSi.jpg	1	
1916	https://pbs.twimg.com/media/CUM2qWaWoAUZO6L.jpg	1	
1915	https://pbs.twimg.com/media/CUM8QZwW4AAVsBl.jpg	1	
1911	https://pbs.twimg.com/media/CUNniSlUYAEj1Jl.jpg	1	
1906	https://pbs.twimg.com/media/CUOGUfJW4AA_eni.jpg	1	
1901	https://pbs.twimg.com/media/CU0a0WXWcAAO_Jy.jpg	1	
1900	https://pbs.twimg.com/media/CU0cVCwWsAERUKY.jpg	1	
1899	https://pbs.twimg.com/media/CUQ7tv3W4AA3KlI.jpg	1	
1895	https://pbs.twimg.com/media/CURwm3cUkAARc06.jpg	1	
1894	https://pbs.twimg.com/media/CUR6jqVWsAEgGot.jpg	1	
1893	https://pbs.twimg.com/media/CUSBemVUEAAAn-6V.jpg	1	
1891	https://pbs.twimg.com/media/CUSEGFNW4AAyyHC.jpg	1	
1920	https://pbs.twimg.com/media/CUL4xR9UkAEdlJ6.jpg	1	
1948	https://pbs.twimg.com/media/CUDmZIkWcAAIPPe.jpg	1	
1945	https://pbs.twimg.com/media/CUELa0NUkAAAscGC.jpg	1	
1944	https://pbs.twimg.com/media/CUES51dXIAEahyG.jpg	1	
1943	https://pbs.twimg.com/media/CUEUva1WsAA2jPb.jpg	1	
1939	https://pbs.twimg.com/media/CUHREBXXAAE6A9b.jpg	1	
1937	https://pbs.twimg.com/media/CUHkkJpXIAA2w3n.jpg	1	
1936	https://pbs.twimg.com/media/CUHr8WbWEAEBPg.jpg	1	
1933	https://pbs.twimg.com/media/CUImtzEVAAAZNJo.jpg	1	
1927	https://pbs.twimg.com/media/CUJJLtWWsAE-go5.jpg	1	
1926	https://pbs.twimg.com/media/CUJK18UWEAEg7AR.jpg	1	
1924	https://pbs.twimg.com/media/CUJUk2iWUAAVt0v.jpg	1	
1923	https://pbs.twimg.com/media/CUJXpRBXIAAN0yz.jpg	1	
1888	https://pbs.twimg.com/media/CUTDtyGXIAARxus.jpg	1	
1883	https://pbs.twimg.com/media/CUT15m1WUAAabZG.jpg	1	
1880	https://pbs.twimg.com/media/CUT9PuQWwAABQv7.jpg	1	
1878	https://pbs.twimg.com/media/CUW37BzWsAAlJlN.jpg	1	
1877	https://pbs.twimg.com/media/CUXDGR2WcAAUQKz.jpg	1	
...	
5219	https://pbs.twimg.com/media/CY-Fn1FWEAQhzhs.jpg	1	
5051	https://pbs.twimg.com/media/CcKC-5LW4AAK-nb.jpg	1	
5048	https://pbs.twimg.com/media/CcMBJODUsAI5-A9.jpg	1	
4771	https://pbs.twimg.com/media/C1B09zOWYAAA1jz.jpg	1	
4774	https://pbs.twimg.com/media/Ck3EribXEAAPhZn.jpg	1	
5488	https://pbs.twimg.com/ext_tw_video_thumb/67574...	1	
5227	https://pbs.twimg.com/media/CY4ciRFUMAAovos.jpg	1	
4290	https://pbs.twimg.com/media/C3CXxaoWQAAiLuC.jpg	1	
4783	https://pbs.twimg.com/media/CkjMx99UoAM2B1a.jpg	1	
5477	https://pbs.twimg.com/media/CWFFt3_XIAArIYK.jpg	1	
5036	https://pbs.twimg.com/media/CcfQgHVWoAAxaay.jpg	1	
5234	https://pbs.twimg.com/media/CYufR8_WQAAWCqo.jpg	1	
5033	https://pbs.twimg.com/media/CcgqBNVW8AE76lv.jpg	1	

5468	https://pbs.twimg.com/media/CWKSIfUUYAAi0B0.jpg	1
4037	https://pbs.twimg.com/ext_tw_video_thumb/88153...	1
5540	https://pbs.twimg.com/media/CVzMPH1UsAELQ_p.jpg	1
5070	https://pbs.twimg.com/media/Cbn40qKWwAADGWt.jpg	1
4249	https://pbs.twimg.com/media/C41st0bXAAE6MP8.jpg	4
5505	https://pbs.twimg.com/media/CV9EvZNUwAAgLCK.jpg	2
5078	https://pbs.twimg.com/media/CbdwATgWwAABGID.jpg	1
5200	https://pbs.twimg.com/media/CZNtgWhWkAAbq3W.jpg	2
5543	https://pbs.twimg.com/media/CVweVUfW4AACPWl.jpg	1
6680	https://pbs.twimg.com/media/CnF8qVDWYAAhOg1.jpg	3
6130	https://pbs.twimg.com/media/C-CYWrvWAAU8AXH.jpg	1
6291	https://pbs.twimg.com/media/C2tugXLXgAArJ04.jpg	1
5994	https://pbs.twimg.com/media/DFi579UWsAAatzw.jpg	1
6116	https://pbs.twimg.com/media/C-3wvtXcAUTuBE.jpg	1
6661	https://pbs.twimg.com/media/Cn7U2xlW8AI9Pqp.jpg	1
6516	https://pbs.twimg.com/media/CtZtJxAXEAAyPGd.jpg	1
6279	https://pbs.twimg.com/media/C3TjvitXAAAI-QH.jpg	1

	p1	p1_conf	...	p3_conf	p3_dog	\
1917	Rottweiler	0.999091	...	0.000157	True	
1916	porcupine	0.989154	...	0.000966	False	
1915	goose	0.980815	...	0.005255	False	
1911	borzoi	0.852088	...	0.005730	False	
1906	hare	0.447893	...	0.088122	True	
1901	toy_poodle	0.787424	...	0.004047	False	
1900	electric_fan	0.984377	...	0.001901	False	
1899	ibex	0.619098	...	0.074673	False	
1895	lorikeet	0.466149	...	0.054247	False	
1894	dalmatian	0.535073	...	0.008164	True	
1893	flat-coated_retriever	0.508392	...	0.048920	True	
1891	miniature_pinscher	0.214200	...	0.104152	True	
1920	teddy	0.709545	...	0.028568	False	
1948	snail	0.999888	...	0.000026	False	
1945	Maltese_dog	0.359383	...	0.106007	True	
1944	Airedale	0.448459	...	0.110183	False	
1943	triceratops	0.442113	...	0.043255	False	
1939	golden_retriever	0.765266	...	0.010667	False	
1937	jigsaw_puzzle	0.560001	...	0.042568	False	
1936	Chihuahua	0.483682	...	0.057495	True	
1933	West_Highland_white_terrier	0.747713	...	0.001804	True	
1927	soft-coated_wheaten_terrier	0.318981	...	0.106014	True	
1926	vizsla	0.259249	...	0.112369	True	
1924	vacuum	0.335830	...	0.140703	False	
1923	Rottweiler	0.283640	...	0.095585	True	
1888	common_iguana	0.999647	...	0.000128	False	
1883	frilled_lizard	0.257695	...	0.085317	False	
1880	hamster	0.172078	...	0.059995	False	
1878	Angora	0.918834	...	0.011015	False	

1877	Arctic_fox	0.473584	...	0.080948	False
...
5219	Pembroke	0.430544	...	0.154352	True
5051	wool	0.525434	...	0.038243	True
5048	doormat	0.201058	...	0.082924	False
4771	kuvasz	0.350629	...	0.087662	True
4774	web_site	0.997154	...	0.000268	False
5488	golden_retriever	0.800495	...	0.068415	True
5227	Pembroke	0.696372	...	0.050592	True
4290	Pembroke	0.990793	...	0.000262	True
4783	golden_retriever	0.843799	...	0.035711	True
5477	Labrador_retriever	0.407430	...	0.074597	True
5036	Siamese_cat	0.202294	...	0.072097	True
5234	Lakeland_terrier	0.873029	...	0.017031	True
5033	pug	0.284428	...	0.138915	False
5468	teddy	0.098283	...	0.077852	False
4037	Samoyed	0.281463	...	0.114854	False
5540	Pomeranian	0.846986	...	0.002605	True
5070	kuvasz	0.805857	...	0.003491	True
4249	briard	0.982755	...	0.004693	True
5505	Pembroke	0.773135	...	0.039036	True
5078	tennis_ball	0.328236	...	0.134080	True
5200	malamute	0.949609	...	0.016663	True
5543	Brabancon_griffon	0.409909	...	0.074820	True
6680	swing	0.999984	...	0.000001	True
6130	flat-coated_retriever	0.321676	...	0.096100	True
6291	Lakeland_terrier	0.196015	...	0.069126	True
5994	Pembroke	0.966327	...	0.004633	True
6116	golden_retriever	0.895529	...	0.019285	True
6661	Airedale	0.602957	...	0.086276	True
6516	padlock	0.731564	...	0.036469	False
6279	Rottweiler	0.681495	...	0.024525	True

	favorites	retweets	user_followers	dog_stage	rating_numerator \
1917	310	85	6992426	None	12
1916	458	248	6992426	None	10
1915	803	604	6992426	None	6
1911	547	235	6992426	None	7.5
1906	1745	1162	6992426	None	12
1901	341	130	6992426	None	9
1900	5961	2386	6993394	None	1
1899	502	332	6992426	None	7
1895	423	253	6992426	None	9
1894	718	341	6992426	None	8
1893	338	98	6992426	None	12
1891	294	65	6992426	None	12
1920	375	168	6992426	None	10
1948	775	496	6992425	None	2

1945	258	98	6992425	None	12
1944	218	80	6992426	None	10
1943	833	566	6992426	None	3
1939	188	51	6992426	None	12
1937	169	49	6992426	None	8
1936	412	126	6992426	None	10
1933	4717	2295	6992426	None	10
1927	624	473	6992426	None	4
1926	193	57	6992426	None	8
1924	759	430	6992426	None	10
1923	396	109	6992426	None	12
1888	649	426	6992426	None	10
1883	510	316	6992427	None	5
1880	1313	831	6992427	None	3
1878	579	300	6992427	None	2
1877	511	330	6992427	None	9
...
5219	2918	954	6992421	pupper	11
5051	2104	607	6992419	pupper	12
5048	2632	761	6993387	pupper	10
4771	6620	2090	6992417	pupper	12
4774	7750	4243	6992417	pupper	14
5488	1217	373	6992427	pupper	12
5227	2466	798	6992422	pupper	12
4290	12692	11557	6992407	pupper	13
4783	10074	3385	6992417	pupper	12
5477	1718	623	6992427	pupper	10
5036	7085	3167	6992419	pupper	13
5234	2678	1043	6992422	pupper	12
5033	2405	597	6992419	pupper	10
5468	2183	590	6992427	pupper	10
4037	49361	16076	6992403	pupper	14
5540	1741	621	6992426	pupper	12
5070	2768	667	6992419	pupper	12
4249	6974	1230	6992406	pupper	13
5505	2915	1387	6992426	pupper	12
5078	2393	723	6992418	pupper	11
5200	2681	807	6992420	pupper	12
5543	1103	381	6992426	pupper	10
6680	7939	3790	6992416	puppo	11
6130	46856	18650	6992403	puppo	13
6291	142980	48957	6992407	puppo	13
5994	48011	10108	6992403	puppo	13
6116	19036	1640	6992403	puppo	13
6661	6941	1697	6992416	puppo	10
6516	23535	8237	6993378	puppo	13
6279	56057	19124	6992407	puppo	14

	rating_denominator	dogs_count	dog_name
1917	10	1	NaN
1916	10	1	NaN
1915	10	1	NaN
1911	10	2	NaN
1906	10	1	NaN
1901	10	1	NaN
1900	10	1	NaN
1899	10	1	NaN
1895	10	1	NaN
1894	10	1	NaN
1893	10	1	NaN
1891	10	1	NaN
1920	10	1	NaN
1948	10	1	NaN
1945	10	1	NaN
1944	10	2	NaN
1943	10	1	NaN
1939	10	1	NaN
1937	10	1	NaN
1936	10	1	NaN
1933	10	1	NaN
1927	10	1	NaN
1926	10	1	NaN
1924	10	1	NaN
1923	10	1	NaN
1888	10	1	NaN
1883	10	1	NaN
1880	10	1	NaN
1878	10	1	NaN
1877	10	1	NaN
...
5219	10	1	NaN
5051	10	1	NaN
5048	10	1	NaN
4771	10	1	NaN
4774	10	1	NaN
5488	10	1	NaN
5227	10	1	NaN
4290	10	1	NaN
4783	10	1	NaN
5477	10	1	NaN
5036	10	1	NaN
5234	10	1	NaN
5033	10	1	NaN
5468	10	1	NaN
4037	10	1	NaN
5540	10	1	NaN

5070	10	1	NaN
4249	10	1	NaN
5505	10	1	NaN
5078	10	1	NaN
5200	10	1	NaN
5543	10	1	NaN
6680	10	1	NaN
6130	10	1	NaN
6291	10	1	NaN
5994	10	1	NaN
6116	10	1	NaN
6661	10	1	NaN
6516	10	1	NaN
6279	10	1	NaN

[625 rows x 25 columns]

```
In [55]: df_master = df_master.drop(['name'], 1)
```

```
list(df_master)
```

```
Out[55]: ['tweet_id',
          'timestamp',
          'source',
          'text',
          'expanded_urls',
          'jpg_url',
          'img_num',
          'p1',
          'p1_conf',
          'p1_dog',
          'p2',
          'p2_conf',
          'p2_dog',
          'p3',
          'p3_conf',
          'p3_dog',
          'favorites',
          'retweets',
          'user_followers',
          'dog_stage',
          'rating_numerator',
          'rating_denominator',
          'dogs_count',
          'dog_name']
```

```
In [56]: male = ['He', 'he', 'him', 'his', "he's", 'himself']
          female = ['She', 'she', 'her', 'hers', 'herself', "she's"]
```

```

dog_gender = []

for text in df_master['text']:
    # Male
    if any(map(lambda v:v in male, text.split())):
        dog_gender.append('male')
    # Female
    elif any(map(lambda v:v in female, text.split())):
        dog_gender.append('female')
    # If group or not specified
    else:
        dog_gender.append('NaN')

len(dog_gender)

df_master['dog_gender'] = dog_gender

# Test
print("dog_gender count \n", df_master.dog_gender.value_counts())

dog_gender count
NaN      1132
male      636
female    226
Name: dog_gender, dtype: int64

In [58]: df_master.loc[df_master['dog_name'] == 'NaN', 'dog_name'] = None
df_master.loc[df_master['dog_gender'] == 'NaN', 'dog_gender'] = None
df_master.loc[df_master['rating_numerator'] == 'NaN', 'rating_numerator'] = 0
df_master.loc[df_master['rating_denominator'] == 'NaN', 'rating_denominator'] = 0

In [59]: df_master.dtypes

Out[59]: tweet_id      int64
timestamp    object
source       object
text         object
expanded_urls object
jpg_url      object
img_num      int64
p1           object
p1_conf      float64
p1_dog       bool
p2           object
p2_conf      float64
p2_dog       bool

```

```

p3                object
p3_conf           float64
p3_dog            bool
favorites         int64
retweets          int64
user_followers    int64
dog_stage         object
rating_numerator  object
rating_denominator object
dogs_count        object
dog_name          object
dog_gender        object
dtype: object

```

```

In [60]: df_master['tweet_id'] = df_master['tweet_id'].astype(object)
df_master['timestamp'] = pd.to_datetime(df_master.timestamp)
df_master['source'] = df_master['source'].astype('category')
df_master['favorites'] = df_master['favorites'].astype(int)
df_master['retweets'] = df_master['retweets'].astype(int)
df_master['user_followers'] = df_master['user_followers'].astype(int)
df_master['dog_stage'] = df_master['dog_stage'].astype('category')
df_master['rating_numerator'] = df_master['rating_numerator'].astype(float)
df_master['rating_denominator'] = df_master['rating_denominator'].astype(float)
df_master['dogs_count'] = df_master['dogs_count'].astype(int)
df_master['dog_gender'] = df_master['dog_gender'].astype('category')

```

```

In [61]: df_master.dtypes

```

```

Out[61]: tweet_id                object
timestamp          datetime64[ns]
source             category
text              object
expanded_urls      object
jpg_url           object
img_num           int64
p1                object
p1_conf           float64
p1_dog            bool
p2               object
p2_conf           float64
p2_dog            bool
p3               object
p3_conf           float64
p3_dog            bool
favorites         int64
retweets          int64
user_followers    int64
dog_stage         category

```

```

rating_numerator      float64
rating_denominator     float64
dogs_count             int64
dog_name               object
dog_gender              category
dtype: object

```

```

In [62]: print(df_master.rating_denominator.value_counts())
df_master.drop('rating_denominator', 1, inplace = True)

```

```

10.0    1981
80.0      2
50.0      2
170.0     1
150.0     1
130.0     1
70.0      1
0.0       1
120.0     1
40.0      1
90.0      1
110.0     1

```

```

Name: rating_denominator, dtype: int64

```

```

In [63]: list(df_master)

```

```

Out[63]: ['tweet_id',
          'timestamp',
          'source',
          'text',
          'expanded_urls',
          'jpg_url',
          'img_num',
          'p1',
          'p1_conf',
          'p1_dog',
          'p2',
          'p2_conf',
          'p2_dog',
          'p3',
          'p3_conf',
          'p3_dog',
          'favorites',
          'retweets',
          'user_followers',
          'dog_stage',
          'rating_numerator',
          'dogs_count',

```

```

'dog_name',
'dog_gender']

```

```
In [64]: df_master.to_csv('twitter_archive_master.csv', index=False, encoding = 'utf-8')
```

```
In [65]: df_master = pd.read_csv('twitter_archive_master.csv')
df_master.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1994 entries, 0 to 1993
Data columns (total 24 columns):
tweet_id          1994 non-null int64
timestamp         1994 non-null object
source            1994 non-null object
text              1994 non-null object
expanded_urls     1994 non-null object
jpg_url           1994 non-null object
img_num           1994 non-null int64
p1                1994 non-null object
p1_conf           1994 non-null float64
p1_dog            1994 non-null bool
p2                1994 non-null object
p2_conf           1994 non-null float64
p2_dog            1994 non-null bool
p3                1994 non-null object
p3_conf           1994 non-null float64
p3_dog            1994 non-null bool
favorites          1994 non-null int64
retweets           1994 non-null int64
user_followers    1994 non-null int64
dog_stage          1994 non-null object
rating_numerator  1994 non-null float64
dogs_count         1994 non-null int64
dog_name           1369 non-null object
dog_gender         862 non-null object
dtypes: bool(3), float64(4), int64(6), object(11)
memory usage: 333.1+ KB

```

```
In [66]: df_master
```

```

Out[66]:

```

	tweet_id	timestamp	source	\
0	667405339315146752	2015-11-19 18:13:27	Twitter for iPhone	
1	667435689202614272	2015-11-19 20:14:03	Twitter for iPhone	
2	667437278097252352	2015-11-19 20:20:22	Twitter for iPhone	
3	667443425659232256	2015-11-19 20:44:47	Twitter for iPhone	
4	667453023279554560	2015-11-19 21:22:56	Twitter Web Client	
5	667455448082227200	2015-11-19 21:32:34	Twitter Web Client	
6	667470559035432960	2015-11-19 22:32:36	Twitter Web Client	

7	667491009379606528	2015-11-19	23:53:52	Twitter Web Client
8	667495797102141441	2015-11-20	00:12:54	Twitter Web Client
9	667502640335572993	2015-11-20	00:40:05	Twitter Web Client
10	667509364010450944	2015-11-20	01:06:48	Twitter Web Client
11	667517642048163840	2015-11-20	01:39:42	Twitter Web Client
12	667524857454854144	2015-11-20	02:08:22	Twitter Web Client
13	667530908589760512	2015-11-20	02:32:25	Twitter Web Client
14	667534815156183040	2015-11-20	02:47:56	Twitter Web Client
15	667538891197542400	2015-11-20	03:04:08	Twitter Web Client
16	667544320556335104	2015-11-20	03:25:43	Twitter Web Client
17	667546741521195010	2015-11-20	03:35:20	Twitter Web Client
18	667549055577362432	2015-11-20	03:44:31	Twitter Web Client
19	667724302356258817	2015-11-20	15:20:54	Twitter Web Client
20	667728196545200128	2015-11-20	15:36:22	Twitter Web Client
21	667766675769573376	2015-11-20	18:09:16	Twitter Web Client
22	667773195014021121	2015-11-20	18:35:10	Twitter Web Client
23	667782464991965184	2015-11-20	19:12:01	Twitter for iPhone
24	667793409583771648	2015-11-20	19:55:30	Twitter for iPhone
25	667801013445750784	2015-11-20	20:25:43	Twitter for iPhone
26	667806454573760512	2015-11-20	20:47:20	Twitter for iPhone
27	667393430834667520	2015-11-19	17:26:08	Twitter for iPhone
28	667832474953625600	2015-11-20	22:30:44	Twitter for iPhone
29	667369227918143488	2015-11-19	15:49:57	Twitter for iPhone
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1971	825026590719483904	2017-01-27	17:04:02	Twitter for iPhone
1972	889531135344209921	2017-07-24	17:02:04	Twitter for iPhone
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1977	793195938047070209	2016-10-31	21:00:23	Twitter for iPhone
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1980	859607811541651456	2017-05-03	03:17:27	Twitter for iPhone
1981	787717603741622272	2016-10-16	18:11:26	Twitter for iPhone
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1983	738537504001953792	2016-06-03	01:07:16	Twitter for iPhone
1984	744995568523612160	2016-06-20	20:49:19	Twitter for iPhone
1985	756275833623502848	2016-07-21	23:53:04	Twitter for iPhone
1986	780931614150983680	2016-09-28	00:46:20	Twitter for iPhone
1987	802239329049477120	2016-11-25	19:55:35	Twitter for iPhone

1988	803773340896923648	2016-11-30 01:31:12	Twitter for iPhone
1989	790946055508652032	2016-10-25 16:00:09	Twitter for iPhone
1990	867421006826221569	2017-05-24 16:44:18	Twitter for iPhone
1991	878776093423087618	2017-06-25 00:45:22	Twitter for iPhone
1992	825535076884762624	2017-01-29 02:44:34	Twitter for iPhone
1993	743253157753532416	2016-06-16 01:25:36	Twitter for iPhone

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                                text \
0   This is Biden. Biden just tripped... 7/10 http...
1       Ermergerd 12/10 https://t.co/PQni2sjPsm
2   Never seen this breed before. Very pointy pup...
3   Exotic dog here. Long neck. Weird paws. Obsess...
4   Meet Cupcake. I would do unspeakable things fo...
5   This is Reese and Twips. Reese protects Twips...
6   This is a northern Wahoo named Kohl. He runs t...
7   Two dogs in this one. Both are rare Jujitsu Py...
8   This is Philippe from Soviet Russia. Commandin...
9   Say hello to Hall and Oates. Oates is winking ...
10  This a Norwegian Pewterschmidt named Tickles. ...
11  This is Dook & Milo. Dook is struggling to...
12  Another topnotch dog. His name is Big Jumpy Ra...
13  Meet Naphaniel. He doesn't necessarily enjoy h...
14  This is Frank (pronounced "Fronq"). Too many b...
15  This is a southwest Coriander named Klint. Hat...
16  This is Kial. Kial is either wearing a cape, w...
17  Here is George. George took a selfie of his ne...
18  Never seen dog like this. Breathes heavy. Tilt...
19  What a dog to start the day with. Very calm. L...
20  Meet Olive. He comes to spot by tree to remini...
21  This is Calvin. He is a Luxembourgian Mayo. Ha...
22  This is a rare Hungarian Pinot named Jessiga. ...
23  Super rare dog. Endangered (?). Thinks it's fu...
24  Dogs only please. Small cows and other non can...
25      OMIGOD 12/10 https://t.co/SVMF4Frflw
26  This is Filup. He is overcome with joy after f...
27  This is Fwed. He is a Canadian Asian Taylormad...
28  THE EYES 12/10\n\nI'm sorry. These are suppose...
29  Here we have a neat pup. Very white. Cool shad...
...
1964 This pupper can only sleep on shoes. It's a cr...
1965 Meet Zuzu. He just graduated college. Astute p...
1966 Say hello to Gizmo. He's quite the pupper. Con...
1967 This is Jed. He may be the fanciest pupper in ...
1968 This pupper just wants a belly rub. This puppe...
1969 This is Sansa. She's gotten too big for her ch...
1970 This is Lucy. She's terrified of the stuffed b...
1971 Say hello to Pablo. He's one gorgeous puppo. A...
1972 This is Stuart. He's sporting his favorite fan...

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1973 Hopefully this puppo on a swing will help get ...
 1974 Here's a puppo participating in the #ScienceMa...
 1975 Here's a super supportive puppo participating ...
 1976 This is Cooper. He's just so damn happy. 10/10...
 1977 Say hello to Lily. She's pupset that her costu...
 1978 This is Sebastian. He can't see all the colors...
 1979 Here's a puppo that seems to be on the fence a...
 1980 Sorry for the lack of posts today. I came home...
 1981 This is Tonks. She is a service puppo. Can hea...
 1982 This is Oliver. He has dreams of being a servi...
 1983 This is Bayley. She fell asleep trying to esca...
 1984 This is Abby. She got her face stuck in a glas...
 1985 When ur older siblings get to play in the deep...
 1986 I want to finally rate this iconic puppo who t...
 1987 This is Loki. He'll do your taxes for you. Can...
 1988 This is Diogi. He fell in the pool as soon as ...
 1989 This is Betty. She's assisting with the dishes...
 1990 This is Shikha. She just watched you drop a sk...
 1991 This is Snoopy. He's a proud #PrideMonthPuppo...
 1992 Here's a very loving and accepting puppo. Appe...
 1993 This is Kilo. He cannot reach the snackum. Nif...

	expanded_urls \
0	https://twitter.com/dog_rates/status/667405339...
1	https://twitter.com/dog_rates/status/667435689...
2	https://twitter.com/dog_rates/status/667437278...
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4	https://twitter.com/dog_rates/status/667453023...
5	https://twitter.com/dog_rates/status/667455448...
6	https://twitter.com/dog_rates/status/667470559...
7	https://twitter.com/dog_rates/status/667491009...
8	https://twitter.com/dog_rates/status/667495797...
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25 https://twitter.com/dog_rates/status/667801013...
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 27 https://twitter.com/dog_rates/status/667393430...
 28 https://twitter.com/dog_rates/status/667832474...
 29 https://twitter.com/dog_rates/status/667369227...
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 1970 https://twitter.com/dog_rates/status/689623661...
 1971 <https://www.gofundme.com/my-puppys-double-cata...>
 1972 https://twitter.com/dog_rates/status/889531135...
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 1980 https://twitter.com/dog_rates/status/859607811...
 1981 https://twitter.com/dog_rates/status/787717603...
 1982 <https://www.gofundme.com/servicedogoliver,http...>
 1983 https://twitter.com/dog_rates/status/738537504...
 1984 https://twitter.com/dog_rates/status/744995568...
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5	https://pbs.twimg.com/media/CUNHMXTU8AAS3HH.jpg	1
6	https://pbs.twimg.com/media/CUNU78YWEAECmpB.jpg	1
7	https://pbs.twimg.com/media/CUNniSlUYAEj1Jl.jpg	1
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0	Saint_Bernard	0.381377	True	...	0.069357
1	Rottweiler	0.999091	True	...	0.000157
2	porcupine	0.989154	False	...	0.000966
3	goose	0.980815	False	...	0.005255
4	Labrador_retriever	0.825670	True	...	0.054018
5	Tibetan_terrier	0.676376	True	...	0.040576
6	toy_poodle	0.304175	True	...	0.073316
7	borzoi	0.852088	True	...	0.005730
8	Chihuahua	0.143957	True	...	0.092482
9	Labrador_retriever	0.996709	True	...	0.000712
10	beagle	0.636169	True	...	0.082549
11	Italian_greyhound	0.125176	True	...	0.081347
12	hare	0.447893	False	...	0.088122
13	golden_retriever	0.633037	True	...	0.046184
14	Pembroke	0.435254	True	...	0.033158
15	Yorkshire_terrier	0.618957	True	...	0.053412
16	Pomeranian	0.412893	True	...	0.071960
17	toy_poodle	0.787424	True	...	0.004047
18	electric_fan	0.984377	False	...	0.001901
19	ibex	0.619098	False	...	0.074673
20	kuvasz	0.360159	True	...	0.270673
21	fire_engine	0.883493	False	...	0.012773
22	West_Highland_white_terrier	0.360465	True	...	0.069038
23	lorikeet	0.466149	False	...	0.054247
24	dalmatian	0.535073	True	...	0.008164
25	flat-coated_retriever	0.508392	True	...	0.048920
26	toyshop	0.253089	False	...	0.112799
27	papillon	0.557009	True	...	0.073473
28	miniature_pinscher	0.214200	True	...	0.104152
29	teddy	0.709545	False	...	0.028568
...
1964	malamute	0.949609	True	...	0.016663
1965	bow	0.168020	False	...	0.043312
1966	Pomeranian	0.943331	True	...	0.007165
1967	cocker_spaniel	0.437216	True	...	0.157402
1968	Brabancon_griffon	0.409909	True	...	0.074820
1969	kelpie	0.219179	True	...	0.074440
1970	toy_poodle	0.279604	True	...	0.077481
1971	Eskimo_dog	0.524454	True	...	0.004976
1972	golden_retriever	0.953442	True	...	0.007958
1973	swing	0.999984	False	...	0.000001
1974	flat-coated_retriever	0.321676	True	...	0.096100
1975	Lakeland_terrier	0.196015	True	...	0.069126

1976	Labrador_retriever	0.929390	True	...	0.007610
1977	Labrador_retriever	0.654762	True	...	0.042339
1978	Cardigan	0.806674	True	...	0.049182
1979	Pembroke	0.966327	True	...	0.004633
1980	golden_retriever	0.895529	True	...	0.019285
1981	German_shepherd	0.992339	True	...	0.000853
1982	American_Staffordshire_terrier	0.925505	True	...	0.020412
1983	chow	0.808737	True	...	0.026498
1984	Old_English_sheepdog	0.427481	True	...	0.134269
1985	Airedale	0.602957	True	...	0.086276
1986	padlock	0.731564	False	...	0.036469
1987	Eskimo_dog	0.482498	True	...	0.134589
1988	miniature_pinscher	0.817066	True	...	0.034195
1989	dishwasher	0.700466	False	...	0.039012
1990	Eskimo_dog	0.616457	True	...	0.001670
1991	Italian_greyhound	0.734684	True	...	0.039725
1992	Rottweiler	0.681495	True	...	0.024525
1993	malamute	0.442612	True	...	0.177822

	p3_dog	favorites	retweets	user_followers	dog_stage	rating_numerator \
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1	True	310	85	6992426	None	12.0
2	False	458	248	6992426	None	10.0
3	False	803	604	6992426	None	6.0
4	True	320	93	6992426	None	11.0
5	True	195	65	6992426	None	7.0
6	True	264	100	6992426	None	11.0
7	False	547	235	6992426	None	7.5
8	False	540	286	6992426	None	9.0
9	True	544	227	6992426	None	11.0
10	True	6956	2210	6992426	None	12.0
11	True	377	198	6992426	None	8.0
12	True	1745	1162	6992426	None	12.0
13	True	487	253	6992426	None	10.0
14	True	841	559	6992426	None	8.0
15	True	209	69	6992426	None	9.0
16	True	889	547	6992426	None	7.0
17	False	341	130	6992426	None	9.0
18	False	5961	2386	6993394	None	1.0
19	False	502	332	6992426	None	7.0
20	True	383	156	6992426	None	11.0
21	False	456	228	6992426	None	9.0
22	False	240	59	6992426	None	8.0
23	False	423	253	6992426	None	9.0
24	True	718	341	6992426	None	8.0
25	True	338	98	6992426	None	12.0
26	True	1084	520	6992426	None	10.0
27	True	199	58	6992426	None	8.0

28	True	294	65	6992426	None	12.0
29	False	375	168	6992426	None	10.0
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1968	True	1103	381	6992426	pupper	10.0
1969	False	11044	3583	6992419	pupper	11.0
1970	True	2391	715	6992421	pupper	10.0
1971	True	6856	1432	6992407	puppo	12.0
1972	True	15058	2243	6992403	puppo	13.0
1973	True	7939	3790	6992416	puppo	11.0
1974	True	46856	18650	6992403	puppo	13.0
1975	True	142980	48957	6992407	puppo	13.0
1976	True	5468	1438	6992416	puppo	10.0
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1986	False	23535	8237	6993378	puppo	13.0
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1989	True	18122	5291	6992411	puppo	12.0
1990	True	16341	2583	6992403	puppo	12.0
1991	True	19371	4155	6992403	puppo	13.0
1992	True	56057	19124	6992407	puppo	14.0
1993	True	4516	1326	6992417	puppo	10.0

	dogs_count	dog_name	dog_gender
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1	1	NaN	NaN
2	1	NaN	NaN
3	1	NaN	NaN
4	1	Cupcake	NaN
5	1	Reese	NaN
6	1	Kohl	male
7	2	NaN	NaN
8	1	Philippe	NaN
9	1	Hall	NaN
10	1	Tickles	NaN
11	1	Dook	male
12	1	NaN	NaN

13	1	Naphaniel	male
14	1	Frank	NaN
15	1	Klint	NaN
16	2	Kial	NaN
17	1	NaN	male
18	1	NaN	NaN
19	1	NaN	NaN
20	1	Olive	male
21	1	Calvin	male
22	1	Jessiga	female
23	1	NaN	NaN
24	1	NaN	NaN
25	1	NaN	NaN
26	1	Filup	male
27	1	Fwed	male
28	1	NaN	NaN
29	1	NaN	NaN
...
1964	1	NaN	male
1965	1	Zuzu	male
1966	1	Gizmo	NaN
1967	1	Jed	male
1968	1	NaN	NaN
1969	1	Sansa	female
1970	1	Lucy	NaN
1971	1	Pablo	NaN
1972	1	Stuart	male
1973	1	NaN	NaN
1974	1	NaN	female
1975	1	NaN	NaN
1976	1	Cooper	NaN
1977	1	Lily	female
1978	1	Sebastian	male
1979	1	NaN	NaN
1980	1	NaN	female
1981	1	Tonks	female
1982	1	Oliver	male
1983	1	Bayley	female
1984	1	Abby	female
1985	1	NaN	NaN
1986	1	NaN	NaN
1987	1	Loki	NaN
1988	1	Diogi	male
1989	1	Betty	NaN
1990	1	Shikha	female
1991	1	Snoopy	NaN
1992	1	NaN	female
1993	1	Kilo	male

```
[1994 rows x 24 columns]
```

3.1 Visualization

```
In [4]: df_master = pd.read_csv('twitter_archive_master.csv')
```

```
In [6]: df_master.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1994 entries, 0 to 1993
Data columns (total 24 columns):
tweet_id          1994 non-null int64
timestamp         1994 non-null object
source            1994 non-null object
text              1994 non-null object
expanded_urls     1994 non-null object
jpg_url           1994 non-null object
img_num           1994 non-null int64
p1                1994 non-null object
p1_conf           1994 non-null float64
p1_dog            1994 non-null bool
p2                1994 non-null object
p2_conf           1994 non-null float64
p2_dog            1994 non-null bool
p3                1994 non-null object
p3_conf           1994 non-null float64
p3_dog            1994 non-null bool
favorites          1994 non-null int64
retweets           1994 non-null int64
user_followers    1994 non-null int64
dog_stage         1994 non-null object
rating_numerator  1994 non-null float64
dogs_count        1994 non-null int64
dog_name          1369 non-null object
dog_gender        862 non-null object
dtypes: bool(3), float64(4), int64(6), object(11)
memory usage: 333.1+ KB
```

```
In [10]: df_master['tweet_id'] = df_master['tweet_id'].astype(object)
df_master['dog_stage'] = df_master['dog_stage'].astype('category')
df_master['dog_gender'] = df_master['dog_gender'].astype('category')
```

```
df_master.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1994 entries, 0 to 1993
Data columns (total 24 columns):
```

```

tweet_id          1994 non-null object
timestamp         1994 non-null object
source            1994 non-null object
text              1994 non-null object
expanded_urls     1994 non-null object
jpg_url           1994 non-null object
img_num           1994 non-null int64
p1                1994 non-null object
p1_conf           1994 non-null float64
p1_dog            1994 non-null bool
p2                1994 non-null object
p2_conf           1994 non-null float64
p2_dog            1994 non-null bool
p3                1994 non-null object
p3_conf           1994 non-null float64
p3_dog            1994 non-null bool
favorites          1994 non-null int64
retweets          1994 non-null int64
user_followers    1994 non-null int64
dog_stage         1994 non-null category
rating_numerator  1994 non-null float64
dogs_count        1994 non-null int64
dog_name          1369 non-null object
dog_gender        862 non-null category
dtypes: bool(3), category(2), float64(4), int64(5), object(10)
memory usage: 306.1+ KB

```

```

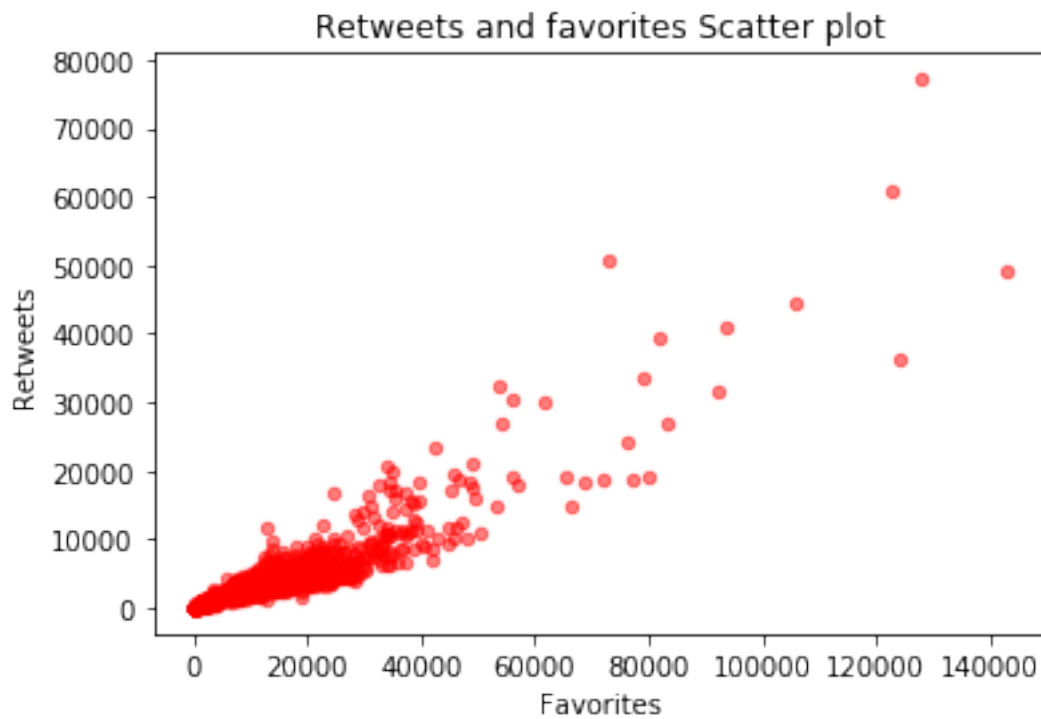
In [11]: df_master.plot(kind = 'scatter', x = 'favorites', y = 'retweets', alpha = 0.5, color =
plt.xlabel('Favorites')
plt.ylabel('Retweets')
plt.title('Retweets and favorites Scatter plot')

```

```

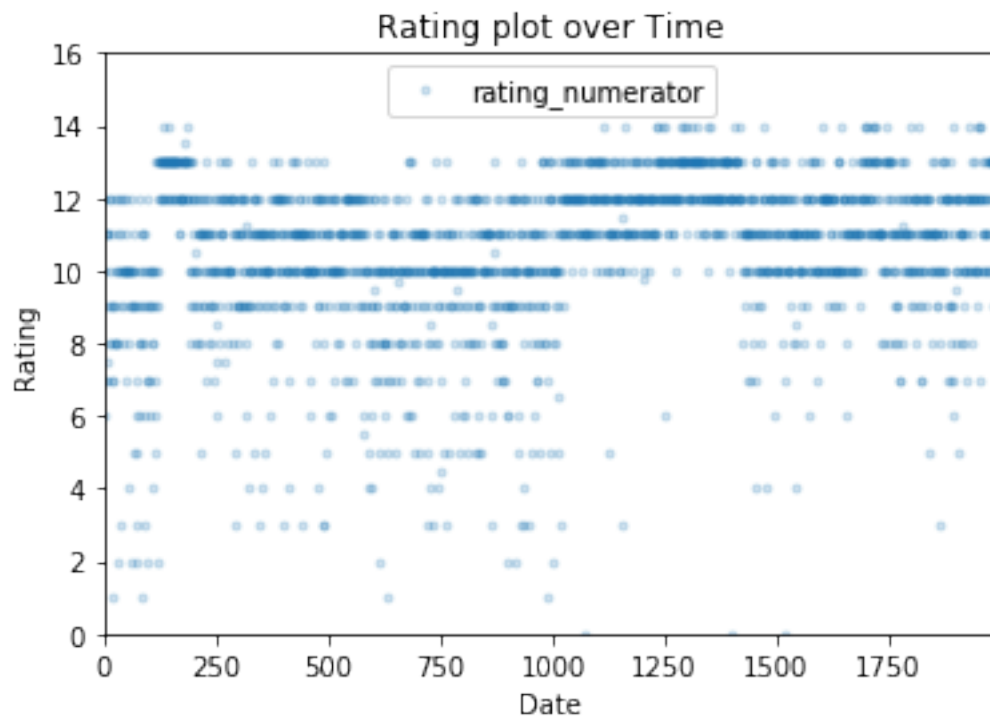
Out[11]: Text(0.5,1,'Retweets and favorites Scatter plot')

```

```
In [14]: df_master.plot(y='rating_numerator', ylim=[0,16], style = '.', alpha = .2)
plt.title('Rating plot over Time')
plt.xlabel('Date')
plt.ylabel('Rating')
```

```
Out[14]: Text(0,0.5,'Rating')
```



```
In [15]: df_master[df_master['dog_gender'].notnull()]['dog_gender'].value_counts().plot(kind = '
plt.title('Dog Gender Partitions')
```

```
Out[15]: Text(0.5,1,'Dog Gender Partitions')
```

